

Exercise 05 Raphael Michel and Florian Stoertz

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1 Structured Learning:

In this exercise we will implement a structured learning system for foreground background segmentation. We will learn the weights of a CRF Potts model.

The first step is to import all needed modules

```
In [10]: # misc
import numpy
import sys

# visualization
import matplotlib.pyplot as plt
import pylab

# features
import skimage.filters
import skimage.feature

# discrete graphical model package
from dgm.models import *
from dgm.solvers import *
from dgm.value_tables import *
# misc. tools
from tools import make_toy_dataset, norm01
import matplotlib.pyplot as plt

from tools import make_toy_dataset, norm01
```

2 The Dataset

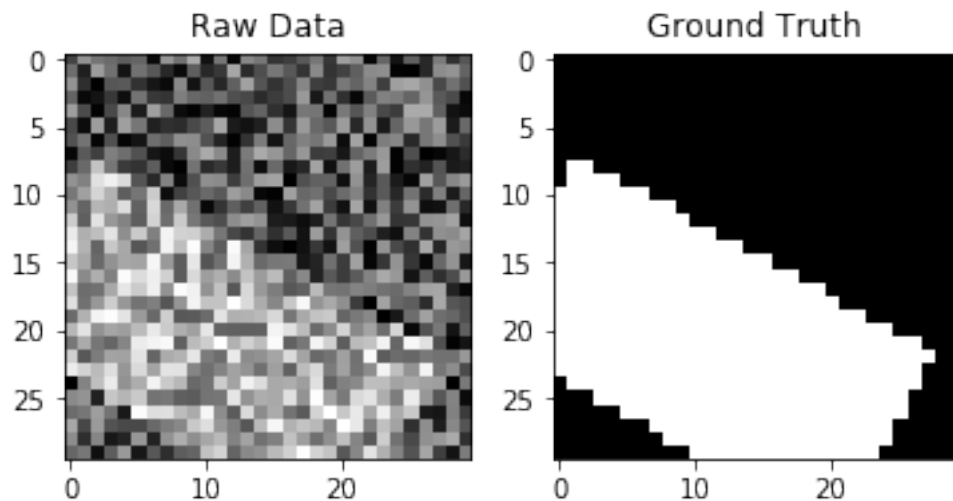
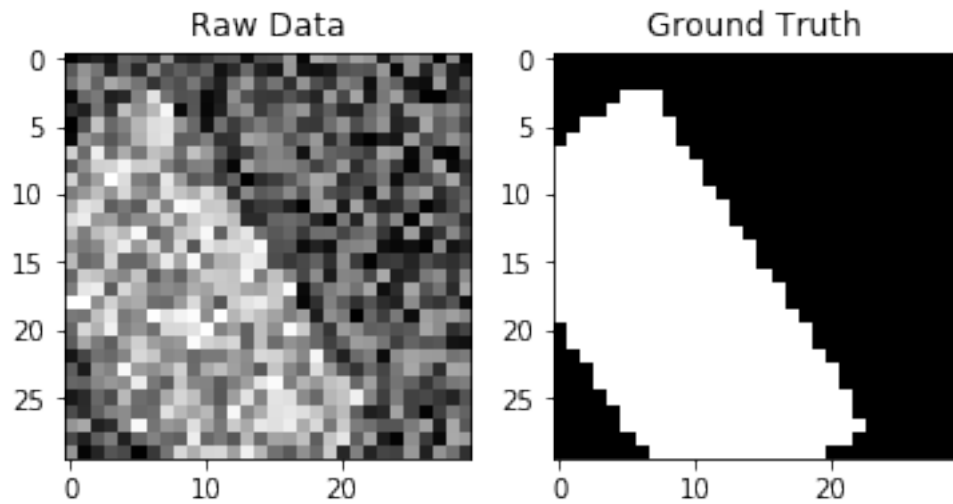
Here we use a toy dataset where we have white squares on black background with some noise. The noise level is given by the variable noise.

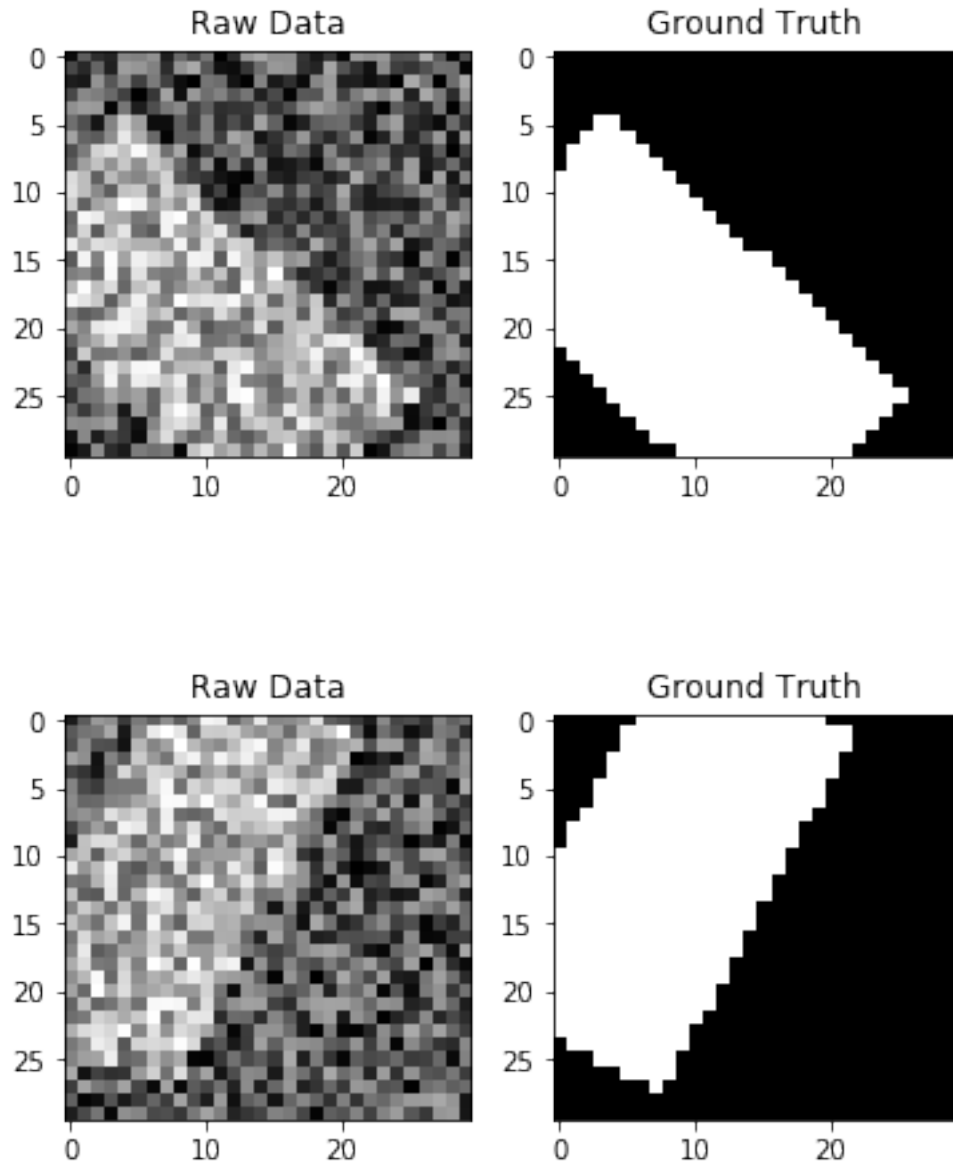
```
In [6]: noise = 2.0
        shape = (30,30)
```

```
x_train, y_train = make_toy_dataset(shape=shape, n_images=5, noise=noise)
x_test , y_test  = make_toy_dataset(shape=shape, n_images=5, noise=noise)
```

```
# show a bit from the dataset
for x in range(4):
    f= pylab.figure()
    ax1 = f.add_subplot(1,2,1)
    pylab.imshow(x_train[x],cmap='gray')
    ax1.set_title('Raw Data')

    ax2 = f.add_subplot(1,2,2)
    pylab.imshow(y_train[x],cmap='gray')
    ax2.set_title('Ground Truth')
    plt.show()
```





3 The Unary Features (5Pt):

Here you need to implement a function which should return pixel wise unary features. You can use features like gaussian smoothing with different sigmas. About 5 features should be enough.

The features should be normalized to be in $[0,1]$

```
In [9]: from skimage.morphology import disk
```

```

def get_unary_features(raw):
    features = []

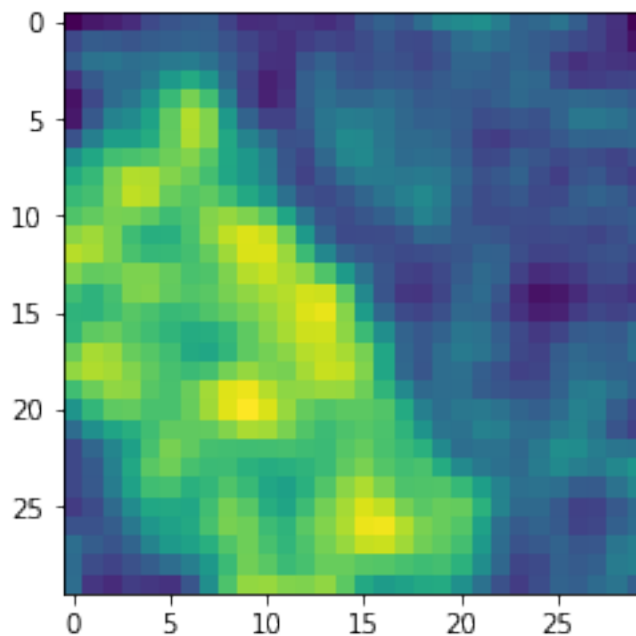
    # Normalize picuteres
    raw -= raw.min()
    raw /= raw.max()

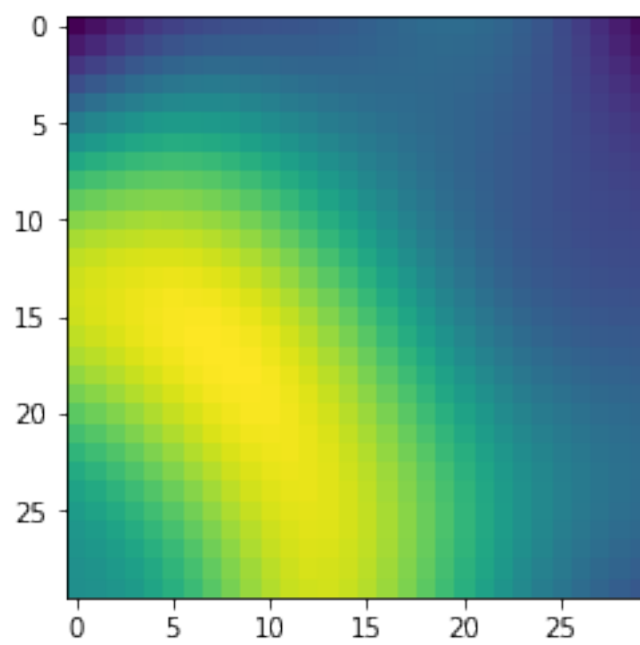
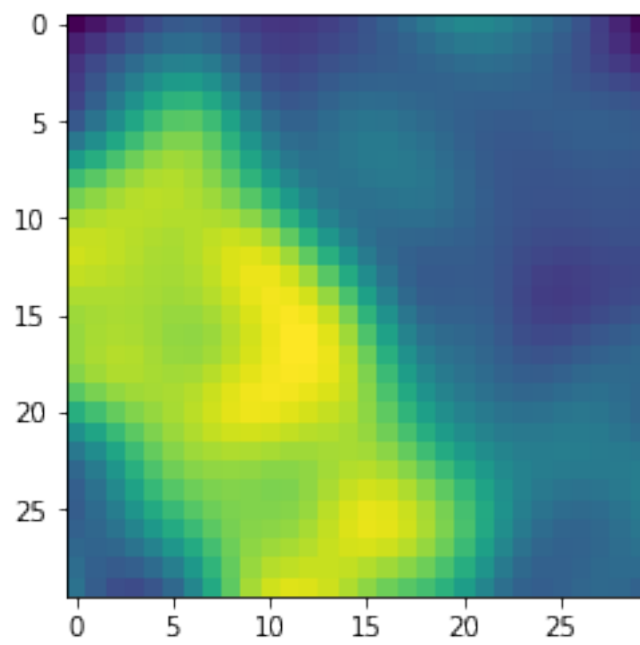
    features.append(skimage.filters.gaussian(raw, sigma=1)[:,:,:None])
    features.append(skimage.filters.gaussian(raw, sigma=2)[:,:,:None])
    features.append(skimage.filters.gaussian(raw, sigma=5)[:,:,:None])
    features.append(skimage.filters.gaussian(raw, sigma=10)[:,:,:None])

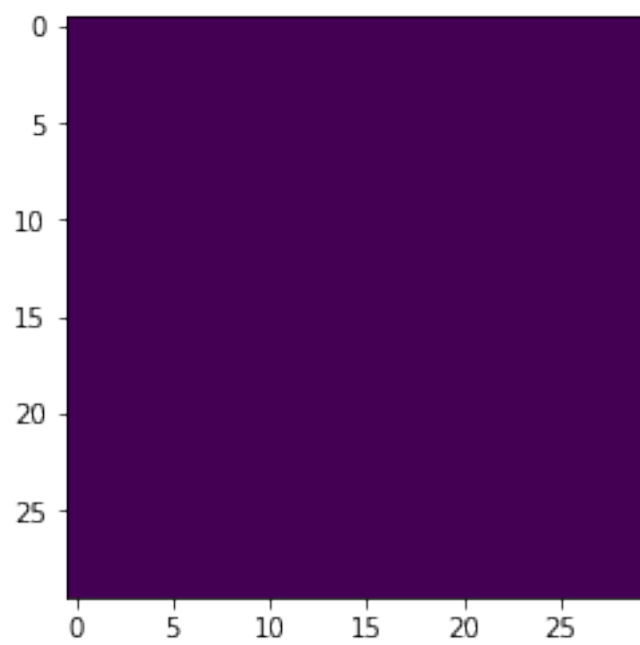
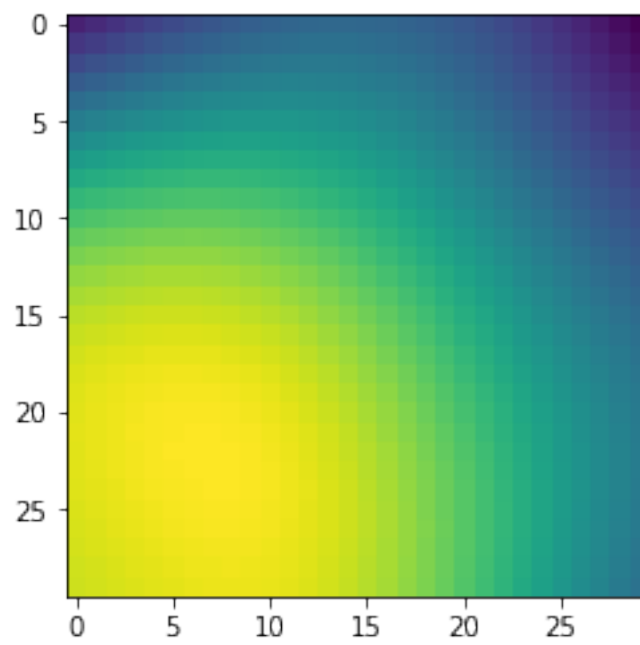
    # Constant feature, keep, because of linearity
    features.append(numpy.ones(raw.shape)[:,:,:None])
    return numpy.concatenate(features, axis=2)

# visualize the features for a raw image
unary_features = get_unary_features(x_train[0])
n_unary_features = unary_features.shape[2]
for i in range(unary_features.shape[2]):
    pylab.imshow(unary_features[:,:,:i])
    pylab.show()

```







4 The Potts Features (5 P):

Here you need to implement a function returning pixel wise features which are used within the potts term. The features should be something like edge detectors.

$e^{-1.0 |edge_strength|}$ should work well.

You can compute edge filters on different sigmas by presmoothing the raw data with a gaussian. About five features should be enough.

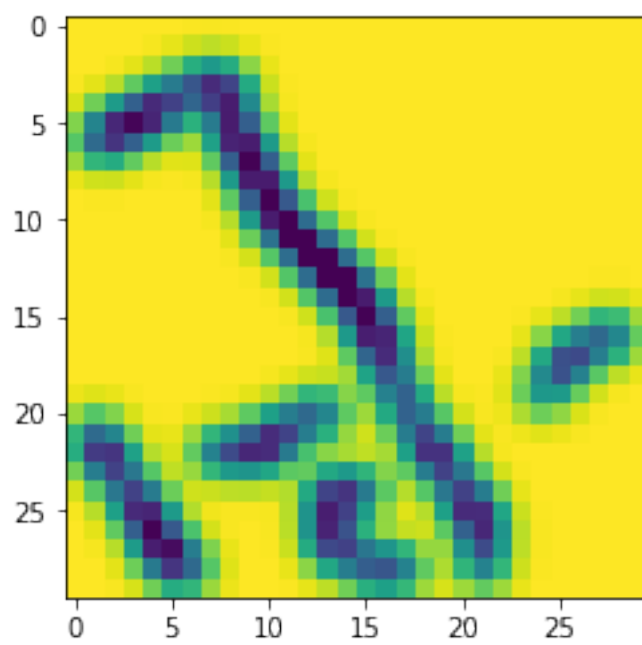
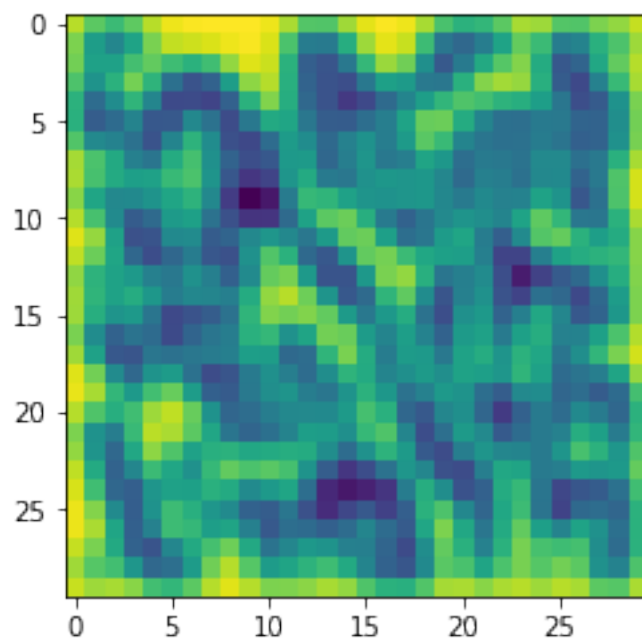
The features should be normalized to be in $[0,1]$

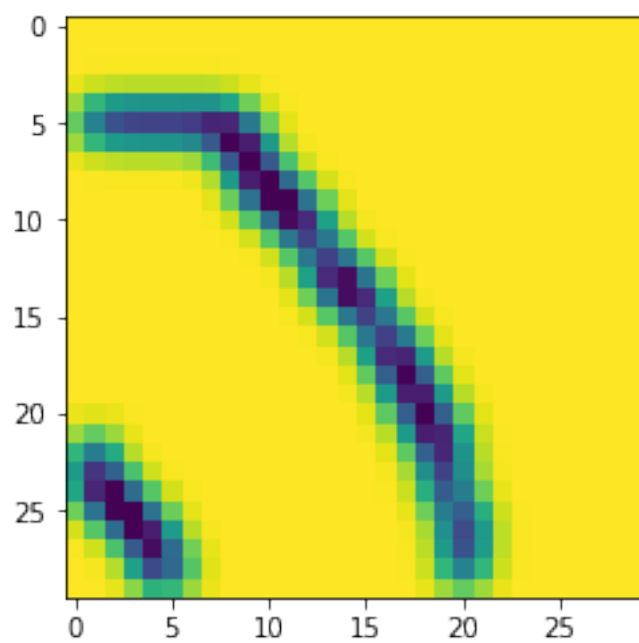
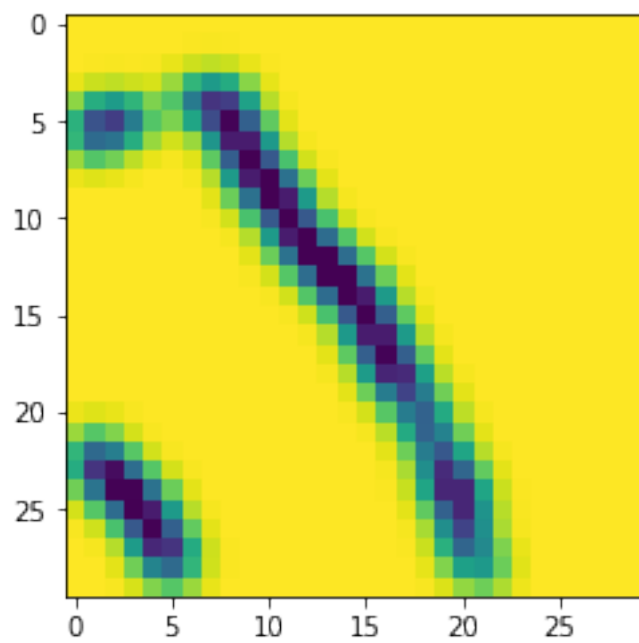
```
In [23]: def get_potts_features(raw):
    features = []

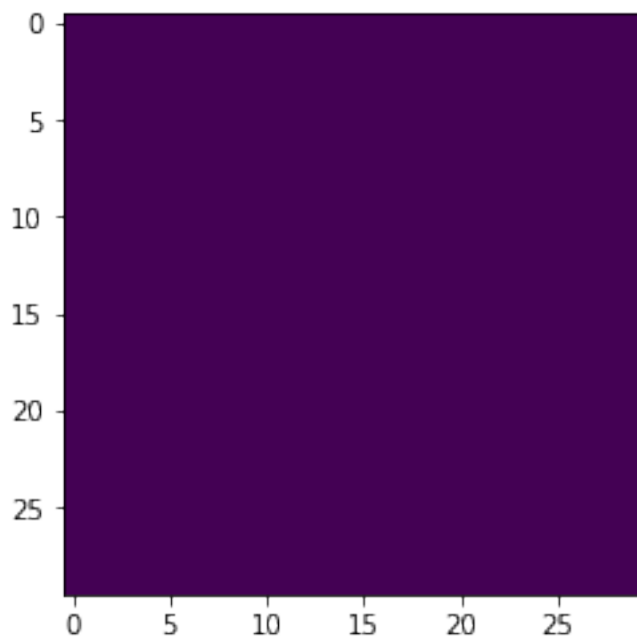
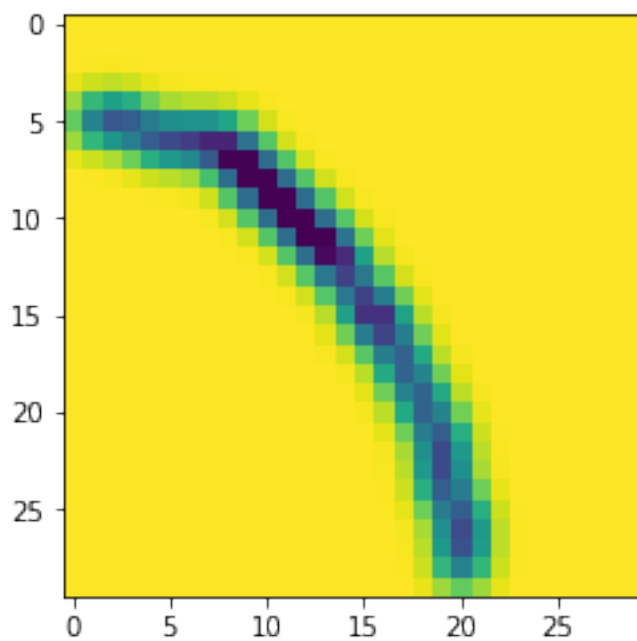
    for sigma in (1, 2, 3, 4, 5):
        edges = skimage.feature.canny(raw, sigma=sigma)
        features.append(
            skimage.filters.gaussian(
                numpy.exp(-1.0 * numpy.abs(edges))
           )[:,:,:None]
        )

    # a constant feature is needed
    features.append(numpy.ones(raw.shape)[:,:,:None])
    return numpy.concatenate(features, axis=2)

# visualize the features for a raw image
potts_features = get_potts_features(x_train[0])
n_potts_features = potts_features.shape[2]
for i in range(potts_features.shape[2]):
    pylab.imshow(potts_features[:,:,:i])
    pylab.show()
```







5 Loss Function:

In this example we will use a very simple `Hamming Loss`.

```
In [24]: class HammingLoss(object):
    def __init__(self, y_true):
        self.y_true = y_true.copy()

    def __call__(self, y_pred):
        """total loss"""
        return numpy.sum(self.y_true!=y_pred)
```

6 Function to set up the weighted Model:

This function will set up the weighted graphical model and also the loss augmented Model:

$$\operatorname{argmin}_y w \cdot \phi(x, y)$$

Loss Augmented Model:

$$\operatorname{argmin}_y w \cdot \phi(x, y) - \Delta(\hat{y}, y)$$

```
In [25]: def build_model(raw_data, gt_image, weights):
    shape = raw_data.shape
    n_var = shape[0] * shape[1]
    n_labels = 2
    variable_space = numpy.ones(n_var)*n_labels

    # lets compute some filters for the unary features
    unary_features = get_unary_features(raw_data)

    # lets compute some filters for the potts features
    potts_features = get_potts_features(raw_data)

    n_weights = potts_features.shape[2] + unary_features.shape[2]

    #print("n_weights",n_weights)
    assert n_weights == len(weights)

    # both graphical models
    gm = WeightedDiscreteGraphicalModel(variable_space=variable_space, weights=weights)
    loss_augmented_gm = WeightedDiscreteGraphicalModel(variable_space=variable_space,

    # convert coordinates to scalar
    def vi(x0,x1): # vi = variable index
        return x1 + x0*shape[1]

    # weight ids for the unaries
    # (just plain numbers to remeber which weights
```

```

# are associated with the unary features)
weight_ids = numpy.arange(unary_features.shape[2])
for x0 in range(shape[0]):
    for x1 in range(shape[1]):

        pixel_val = raw_data[x0, x1]
        gt_label = gt_image[x0, x1]
        features = unary_features[x0, x1, :]

        unary_function = WeightedTwoClassUnary(features=features, weight_ids=weight_ids,
                                                weights=weights)

        if gt_label == 0:
            loss = numpy.array([0,1])
        else:
            loss = numpy.array([1,0])

        loss_augmented_unary_function = WeightedTwoClassUnary(features=features, weight_ids=weight_ids,
                                                                weights=weights, const_terms=-1.0*pixel_val)

        variables = vi(x0,x1)
        gm.add_factor(variables=variables, value_table=unary_function)
        loss_augmented_gm.add_factor(variables=variables, value_table=loss_augmented_unary_function)

# add pairwise factors
# the weight id's for the pairwise factors

# average over 2 coordinates to extract
# extract feature vectors for potts functions
def get_potts_feature_vec(coord_a, coord_b):

    fa = potts_features[coord_a[0],coord_a[1],:]
    fb = potts_features[coord_b[0],coord_b[1],:]
    return (fa+fb)/2.0

# weight ids for the potts functions
# (just plain numbers to remeber which weights
# are associated with the potts features)
weight_ids = numpy.arange(potts_features.shape[2]) + unary_features.shape[2]

for x0 in range(shape[0]):
    for x1 in range(shape[1]):

        # horizontal edge
        if x0 + 1 < shape[0]:
            variables = [vi(x0,x1),vi(x0+1,x1)]
            features = get_potts_feature_vec((x0,x1), (x0+1,x1))
            # the weighted potts function

```

```

potts_function = WeightedPottsFunction(shape=[2,2],
                                       features=features,
                                       weight_ids=weight_ids,
                                       weights=weights)

# add factors to both models
gm.add_factor(variables=variables, value_table=potts_function)
loss_augmented_gm.add_factor(variables=variables, value_table=potts_f

# vertical edge
if x1 + 1 < shape[1]:
    variables = [vi(x0,x1),vi(x0, x1+1)]
    features = get_potts_feature_vec((x0,x1), (x0,x1+1))
    # the weighted potts function
    potts_function = WeightedPottsFunction(shape=[2,2],
                                           features=features,
                                           weight_ids=weight_ids,
                                           weights=weights)

    # add factors to both models
    gm.add_factor(variables=variables, value_table=potts_function)
    loss_augmented_gm.add_factor(variables=variables, value_table=potts_f

# gm, loss augmented and the loss
return gm, loss_augmented_gm, HammingLoss(gt_image.ravel())

```

7 Build the weighted models:

```

In [27]: #
n_weights = n_unary_features + n_potts_features
weights = numpy.zeros(n_weights)

# build the graphical models
models_train = [build_model(x,y, weights) for x,y in zip(x_train, y_train)]
models_test  = [build_model(x,y, weights) for x,y in zip(x_test, y_test)]

# very simple helper class to combine things
class Dataset(object):
    def __init__(self, models_train, models_test, weights):
        self.models_train = models_train
        self.models_test = models_test
        self.weights = weights
dset = Dataset(models_train, models_test, weights)

```

8 Subgradient SSVM

Instead of a cutting plane approach, we use a subgradient decent to find the optimal weights

[Learn more about subgradient svm](#)

```

In [68]: def subgradient_ssvm(dataset, n_iter=20, learning_rate=1.0, c=0.5, lower_bounds=None,

weights = dataset.weights
n = len(dataset.models_train)

if lower_bounds is None:
    lower_bounds = numpy.ones(len(weights))*-1.0*float('inf')

if upper_bounds is None:
    upper_bounds = numpy.ones(len(weights))*float('inf')

do_opt = True
for iteration in range(n_iter):

    effective_learning_rate = learning_rate*float(learning_rate)/(1.0+iteration)

    # compute gradient
    diff = numpy.zeros(weights.shape)
    for gm, gm_loss_augmented, loss_function in dataset.models_train:

        # update the weights to the current weight vector
        gm.change_weights(weights)
        gm_loss_augmented.change_weights(weights)

        # the gt vector
        y_true = loss_function.y_true

        # optimize loss augmented /
        # find most violated constraint
        graphcut = algorithm(model=gm_loss_augmented)
        y_hat = graphcut.optimize()

        # compute joint feature vector
        phi_y_hat = gm.phi(y_hat)
        phi_y_true = gm.phi(y_true)

        diff += phi_y_true - phi_y_hat

    new_weights = weights - effective_learning_rate*(c/n)*diff

    # project new weights
    where_to_large = numpy.where(new_weights>upper_bounds)
    new_weights[where_to_large] = upper_bounds[where_to_large]
    where_to_small = numpy.where(new_weights<lower_bounds)

```

```

new_weights[where_to_small] = lower_bounds[where_to_small]

delta = numpy.abs(new_weights-weights).sum()
if(delta<convergence):
    print("converged")
    break
print('iter',iteration, 'delta',delta," ",numpy.round(new_weights,3))

weights = new_weights

return weights

```

9 Learn The Weights:

We call use graphcut to find the argmin of the gm and the loss augmented gm. Graphcut will find optimal values iff the potts regularizer is positive. To ensure a positive regularizer is learned, we can constraint the weights of the potts function to be positive.

```
In [29]: lower_bounds = numpy.ones(len(weights))*(-1.0*float('inf'))
```

```
# we want the regularizer 'beta' to be positive
```

```
lower_bounds[n_unary_features:n_unary_features+n_potts_features] = 0
```

```
weights = subgradient_ssvm(dset,c=0.5,learning_rate=1.0, lower_bounds=lower_bounds, n
```

```

iter 0 delta 148.301948269    [-32.05 -25.162 -7.94  11.149  72.      0.      0.      0.
  0.   ]
iter 1 delta 315.881335136    [-93.278 -84.535 -62.031 -35.541 -22.5      0.      0.      0.
  0.   ]
iter 2 delta 216.586094929    [-63.143 -53.341 -28.618 -0.698  64.5      0.      0.      0.
  0.   ]
iter 3 delta 15.1156490128    [-61.085 -51.222 -26.47    1.362  68.925  0.4      0.572  0.475
  0.37   0.125]
iter 4 delta 5.8495989785     [-61.669 -51.795 -27.047  0.812  67.665  0.      0.      0.
  0.   ]
iter 5 delta 0.770754366329    [-61.523 -51.65  -26.91    0.938  67.882  0.      0.      0.
  0.   ]
iter 6 delta 0.131381588897    [-61.531 -51.658 -26.923  0.923  67.796  0.      0.      0.
  0.   ]
iter 7 delta 0.069690615693    [-61.52  -51.647 -26.917  0.926  67.758  0.      0.      0.
  0.   ]
iter 8 delta 0.0619472139493    [-61.511 -51.636 -26.911  0.928  67.725  0.      0.      0.
  0.   ]
iter 9 delta 0.0835031517545    [-61.488 -51.612 -26.891  0.945  67.725  0.      0.      0.
  0.   ]
iter 10 delta 0.0506840841403    [-61.48  -51.603 -26.886  0.947  67.698  0.      0.      0.

```

0.]								
iter 11	delta 0.046460410462	[-61.473	-51.596	-26.881	0.949	67.673	0.	0.	0.
0.]								
iter 12	delta 0.0642331936573	[-61.455	-51.577	-26.866	0.961	67.673	0.	0.	0.
0.]								
iter 13	delta 0.0398232089674	[-61.449	-51.57	-26.862	0.963	67.651	0.	0.	0.
0.]								
iter 14	delta 0.0371683283696	[-61.443	-51.564	-26.859	0.965	67.631	0.	0.	0.
0.]								
iter 15	delta 0.0461726220169	[-61.432	-51.552	-26.849	0.972	67.625	0.	0.	0.
0.]								
iter 16	delta 0.0327955838555	[-61.427	-51.547	-26.846	0.974	67.608	0.	0.	0.
0.]								
iter 17	delta 0.0359898123466	[-61.419	-51.539	-26.841	0.977	67.596	0.	0.	0.
0.]								
iter 18	delta 0.0340956116968	[-61.412	-51.531	-26.835	0.981	67.586	0.	0.	0.
0.]								
iter 19	delta 0.0278762462772	[-61.408	-51.526	-26.832	0.982	67.571	0.	0.	0.
0.]								
iter 20	delta 0.0397634055974	[-61.397	-51.515	-26.823	0.99	67.571	0.	0.	0.
0.]								
iter 21	delta 0.0253420420702	[-61.393	-51.511	-26.821	0.991	67.557	0.	0.	0.
0.]								
iter 22	delta 0.0242402141541	[-61.39	-51.507	-26.818	0.992	67.544	0.	0.	0.
0.]								
iter 23	delta 0.0307817480113	[-61.382	-51.499	-26.812	0.997	67.54	0.	0.	0.
0.]								
iter 24	delta 0.0223009970218	[-61.379	-51.495	-26.81	0.998	67.528	0.	0.	0.
0.]								
iter 25	delta 0.0284139212412	[-61.371	-51.488	-26.804	1.002	67.524	0.	0.	0.
0.]								
iter 26	delta 0.0206490713164	[-61.368	-51.484	-26.802	1.003	67.513	0.	0.	0.
0.]								
iter 27	delta 0.0231363079371	[-61.364	-51.479	-26.798	1.006	67.506	0.	0.	0.
0.]								
iter 28	delta 0.0223385042151	[-61.359	-51.474	-26.795	1.008	67.499	0.	0.	0.
0.]								
iter 29	delta 0.0215938874079	[-61.355	-51.469	-26.791	1.011	67.492	0.	0.	0.
0.]								
iter 30	delta 0.0208973103948	[-61.35	-51.465	-26.788	1.013	67.486	0.	0.	0.
0.]								
iter 31	delta 0.0202442694449	[-61.346	-51.46	-26.785	1.015	67.48	0.	0.	0.
0.]								
iter 32	delta 0.0196308067345	[-61.342	-51.456	-26.782	1.017	67.474	0.	0.	0.
0.]								
iter 33	delta 0.0163977919278	[-61.34	-51.453	-26.78	1.018	67.465	0.	0.	0.
0.]								
iter 34	delta 0.0211074843506	[-61.335	-51.448	-26.776	1.021	67.462	0.	0.	0.

0.]								
iter 35 delta 0.0154868034873	[-61.332 -51.445 -26.774	1.022	67.454	0.	0.	0.		
0.]								
iter 36 delta 0.0175085573578	[-61.329 -51.441 -26.772	1.024	67.448	0.	0.	0.		
0.]								
iter 37 delta 0.0170478058484	[-61.325 -51.437 -26.769	1.026	67.443	0.	0.	0.		
0.]								
iter 38 delta 0.0166106826215	[-61.322 -51.434 -26.766	1.027	67.438	0.	0.	0.		
0.]								
iter 39 delta 0.016195415556	[-61.319 -51.43 -26.764	1.029	67.433	0.	0.	0.		
0.]								
iter 40 delta 0.0158004054204	[-61.315 -51.427 -26.761	1.031	67.428	0.	0.	0.		
0.]								
iter 41 delta 0.0154242052914	[-61.312 -51.423 -26.759	1.033	67.423	0.	0.	0.		
0.]								
iter 42 delta 0.0150655028427	[-61.309 -51.42 -26.756	1.034	67.419	0.	0.	0.		
0.]								
iter 43 delta 0.0147231050509	[-61.306 -51.417 -26.754	1.036	67.414	0.	0.	0.		
0.]								
iter 44 delta 0.0143959249386	[-61.303 -51.414 -26.752	1.037	67.41	0.	0.	0.		
0.]								
iter 45 delta 0.0140829700487	[-61.3 -51.41 -26.749	1.039	67.405	0.	0.	0.		
0.]								
iter 46 delta 0.0137833323881	[-61.298 -51.407 -26.747	1.04	67.401	0.	0.	0.		
0.]								
iter 47 delta 0.01349617963	[-61.295 -51.404 -26.745	1.042	67.397	0.	0.	0.		
0.]								
iter 48 delta 0.011378059705	[-61.293 -51.402 -26.744	1.042	67.391	0.	0.	0.		
0.]								
iter 49 delta 0.0129563324448	[-61.29 -51.4 -26.742	1.044	67.387	0.	0.	0.		
0.]								
iter 50 delta 0.0127022867106	[-61.288 -51.397 -26.74	1.045	67.383	0.	0.	0.		
0.]								
iter 51 delta 0.0124580119661	[-61.285 -51.394 -26.738	1.046	67.379	0.	0.	0.		
0.]								
iter 52 delta 0.0122229551366	[-61.283 -51.391 -26.736	1.048	67.375	0.	0.	0.		
0.]								
iter 53 delta 0.0119966041155	[-61.28 -51.389 -26.734	1.049	67.371	0.	0.	0.		
0.]								
iter 54 delta 0.0117784840407	[-61.278 -51.386 -26.732	1.05	67.368	0.	0.	0.		
0.]								
iter 55 delta 0.0115681539685	[-61.276 -51.384 -26.73	1.052	67.364	0.	0.	0.		
0.]								
iter 56 delta 0.0113652038989	[-61.273 -51.381 -26.729	1.053	67.361	0.	0.	0.		
0.]								
iter 57 delta 0.0111692521076	[-61.271 -51.379 -26.727	1.054	67.357	0.	0.	0.		
0.]								
iter 58 delta 0.0109799427498	[-61.269 -51.376 -26.725	1.055	67.354	0.	0.	0.		

0.]								
iter 59	delta 0.010796943704	[-61.267 -51.374 -26.723	1.056	67.35	0.	0.	0.		
0.]								
iter 60	delta 0.0106199446269	[-61.264 -51.371 -26.722	1.058	67.347	0.	0.	0.		
0.]								
iter 61	delta 0.0104486551974	[-61.262 -51.369 -26.72	1.059	67.344	0.	0.	0.		
0.]								
iter 62	delta 0.0102828035276	[-61.26 -51.367 -26.718	1.06	67.341	0.	0.	0.		
0.]								
iter 63	delta 0.0101221347225	[-61.258 -51.365 -26.717	1.061	67.338	0.	0.	0.		
0.]								
iter 64	delta 0.00857730654683	[-61.257 -51.363 -26.716	1.061	67.333	0.	0.	0.		
0.]								
iter 65	delta 0.00981540336725	[-61.255 -51.361 -26.714	1.062	67.33	0.	0.	0.		
0.]								
iter 66	delta 0.00966890480953	[-61.253 -51.359 -26.713	1.063	67.327	0.	0.	0.		
0.]								
iter 67	delta 0.00952671503291	[-61.251 -51.357 -26.711	1.064	67.324	0.	0.	0.		
0.]								
iter 68	delta 0.00938864669911	[-61.249 -51.355 -26.71	1.065	67.321	0.	0.	0.		
0.]								
iter 69	delta 0.00925452317483	[-61.247 -51.353 -26.708	1.066	67.318	0.	0.	0.		
0.]								
iter 70	delta 0.00912417777801	[-61.245 -51.351 -26.707	1.067	67.316	0.	0.	0.		
0.]								
iter 71	delta 0.00899745308665	[-61.244 -51.349 -26.705	1.068	67.313	0.	0.	0.		
0.]								
iter 72	delta 0.00887420030464	[-61.242 -51.347 -26.704	1.069	67.31	0.	0.	0.		
0.]								
iter 73	delta 0.00875427867891	[-61.24 -51.345 -26.703	1.07	67.307	0.	0.	0.		
0.]								
iter 74	delta 0.00863755496318	[-61.238 -51.343 -26.701	1.071	67.305	0.	0.	0.		
0.]								
iter 75	delta 0.0085239029242	[-61.236 -51.341 -26.7	1.072	67.302	0.	0.	0.		
0.]								
iter 76	delta 0.00841320288621	[-61.235 -51.339 -26.698	1.073	67.299	0.	0.	0.		
0.]								
iter 77	delta 0.00830534131075	[-61.233 -51.337 -26.697	1.074	67.297	0.	0.	0.		
0.]								
iter 78	delta 0.00820021040808	[-61.231 -51.335 -26.696	1.075	67.294	0.	0.	0.		
0.]								
iter 79	delta 0.0069690615693	[-61.23 -51.334 -26.695	1.075	67.291	0.	0.	0.		
0.]								
iter 80	delta 0.00799773607702	[-61.229 -51.333 -26.694	1.076	67.288	0.	0.	0.		
0.]								
iter 81	delta 0.00790020271022	[-61.227 -51.331 -26.693	1.077	67.286	0.	0.	0.		
0.]								
iter 82	delta 0.00780501954504	[-61.226 -51.329 -26.691	1.078	67.283	0.	0.	0.		

```

0.  ]
iter 83 delta 0.00771210264569 [-61.224 -51.327 -26.69 1.079 67.281 0. 0. 0
0.  ]
iter 84 delta 0.00762137202633 [-61.222 -51.326 -26.689 1.079 67.279 0. 0. 0
0.  ]
iter 85 delta 0.00753275142138 [-61.221 -51.324 -26.688 1.08 67.276 0. 0. 0
0.  ]
iter 86 delta 0.00744616807172 [-61.219 -51.322 -26.687 1.081 67.274 0. 0. 0
0.  ]
iter 87 delta 0.00736155252543 [-61.218 -51.321 -26.685 1.082 67.272 0. 0. 0
0.  ]
iter 88 delta 0.00727883845212 [-61.216 -51.319 -26.684 1.083 67.269 0. 0. 0
0.  ]
iter 89 delta 0.00719796246931 [-61.215 -51.318 -26.683 1.083 67.267 0. 0. 0
0.  ]
iter 90 delta 0.00711886398064 [-61.213 -51.316 -26.682 1.084 67.265 0. 0. 0
0.  ]
iter 91 delta 0.00704148502433 [-61.212 -51.314 -26.681 1.085 67.263 0. 0. 0
0.  ]
iter 92 delta 0.00696577013159 [-61.211 -51.313 -26.68 1.086 67.261 0. 0. 0
0.  ]
iter 93 delta 0.00593111622918 [-61.21 -51.312 -26.679 1.086 67.257 0. 0. 0
0.  ]
iter 94 delta 0.00681912233936 [-61.208 -51.31 -26.678 1.087 67.255 0. 0. 0
0.  ]
iter 95 delta 0.00674808981498 [-61.207 -51.309 -26.677 1.087 67.253 0. 0. 0
0.  ]
iter 96 delta 0.00667852187875 [-61.206 -51.307 -26.676 1.088 67.251 0. 0. 0
0.  ]
iter 97 delta 0.00661037369631 [-61.204 -51.306 -26.675 1.089 67.249 0. 0. 0
0.  ]
iter 98 delta 0.00654360224483 [-61.203 -51.305 -26.674 1.089 67.247 0. 0. 0
0.  ]
iter 99 delta 0.00647816622238 [-61.202 -51.303 -26.673 1.09 67.245 0. 0. 0
0.  ]

```

10 Training Set Performance:

```

In [30]: print("learnedn weights",weights)
         for i,(gm,_,loss_function) in enumerate(models_train):
             gm.change_weights(weights)

         graphcut = GraphCut(model=gm)
         y_pred = graphcut.optimize()

```

```

prediction_image = y_pred.reshape(shape)

# show a bit from the dataset

f= pylab.figure()
ax1 = f.add_subplot(1,3,1)
pylab.imshow(x_train[i],cmap='gray')
ax1.set_title('Raw Data')

ax2 = f.add_subplot(1,3,2)
pylab.imshow(y_train[i],cmap='gray')
ax2.set_title('Ground Truth')

ax3 = f.add_subplot(1,3,3)
pylab.imshow(prediction_image,cmap='gray')
ax3.set_title('Prediction')

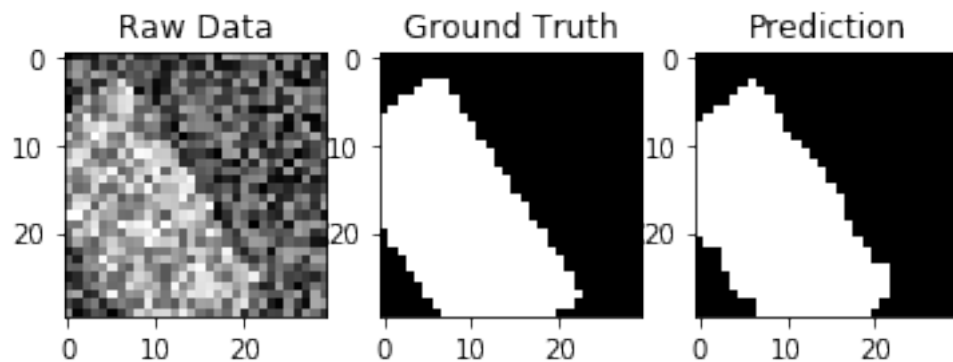
plt.show()

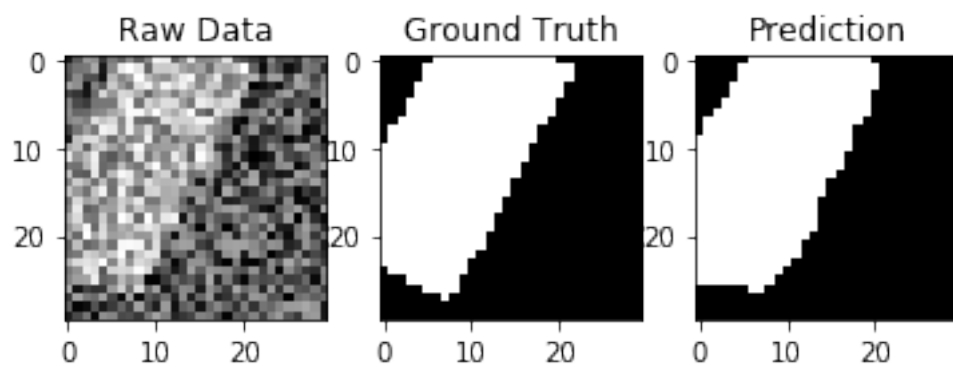
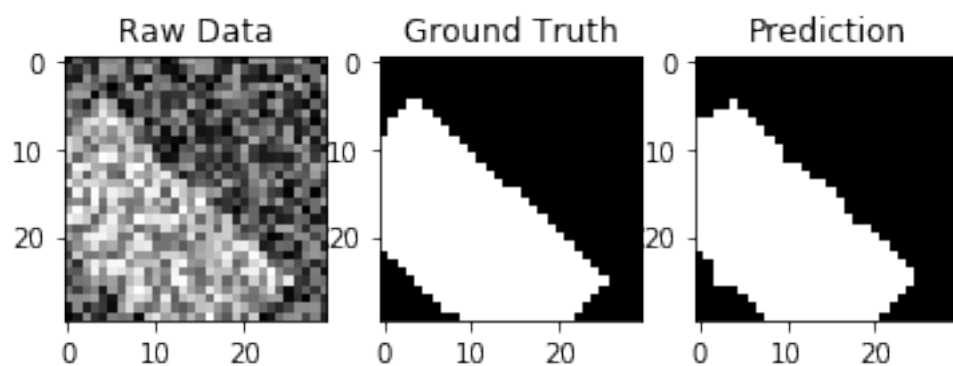
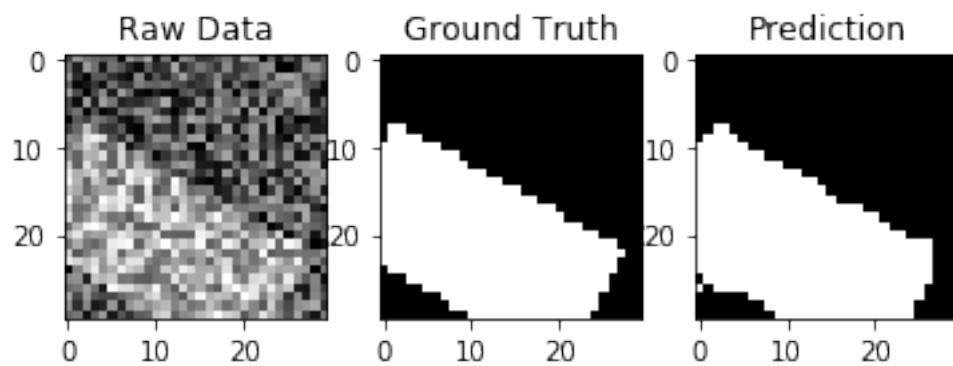
```

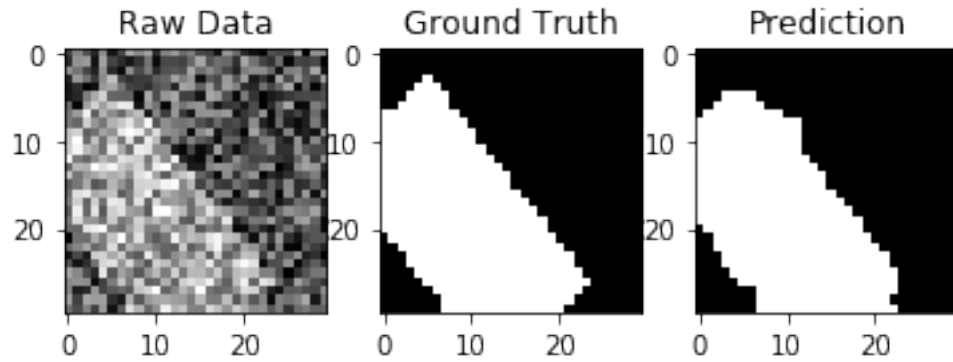
```

learnedn weights [-61.20162564 -51.30311871 -26.67284331  1.09018464  67.24514042  0.
0.          0.          0.          0.          ]

```







11 Test set performance:

```
In [31]: for i,(gm,_,loss_function) in enumerate(models_test):
          gm.change_weights(weights)
```

```
graphcut = GraphCut(model=gm)
y_pred = graphcut.optimize()
```

```
prediction_image = y_pred.reshape(shape)
```

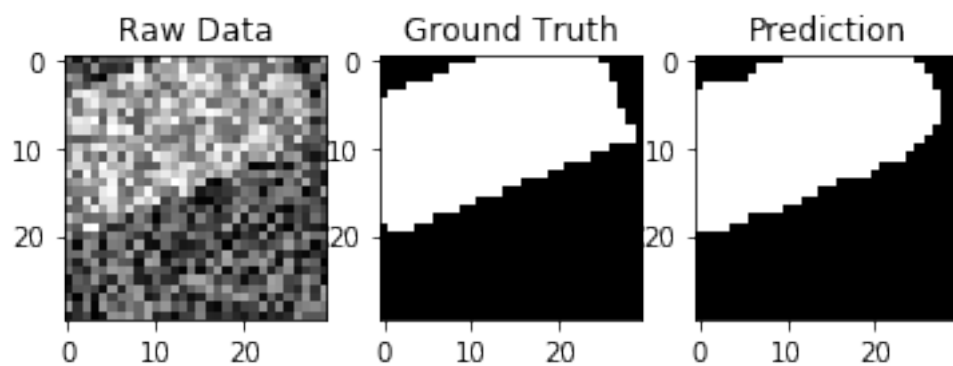
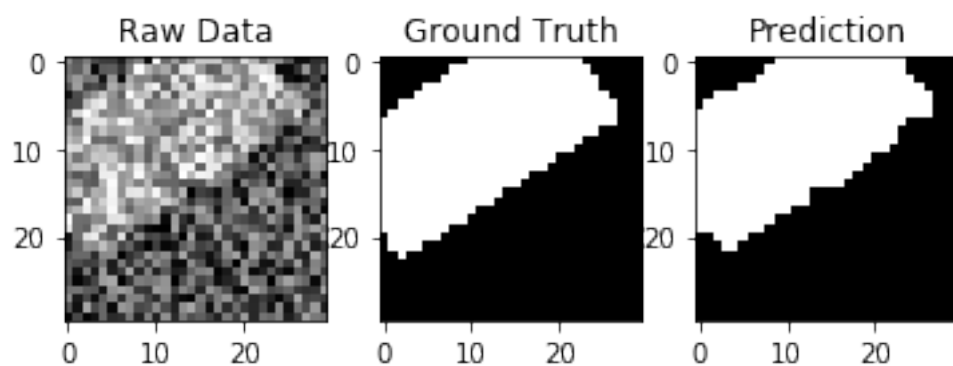
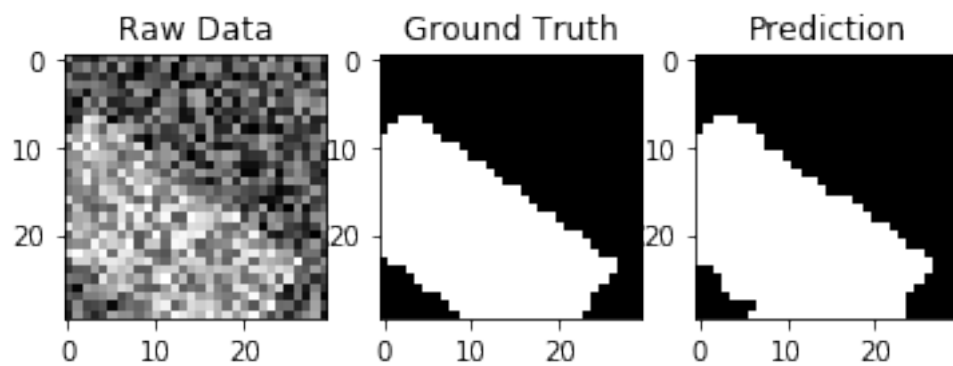
```
# show a bit from the dataset
```

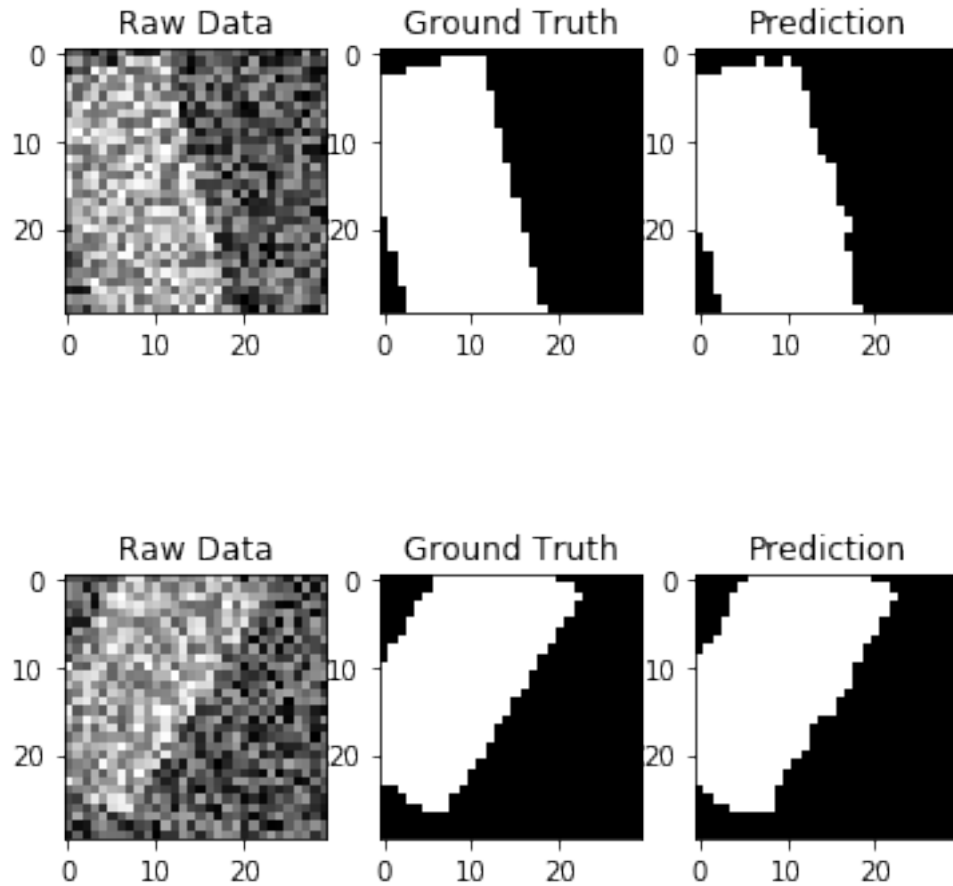
```
f= pylab.figure()
ax1 = f.add_subplot(1,3,1)
pylab.imshow(x_test[i],cmap='gray')
ax1.set_title('Raw Data')
```

```
ax2 = f.add_subplot(1,3,2)
pylab.imshow(y_test[i],cmap='gray')
ax2.set_title('Ground Truth')
```

```
ax3 = f.add_subplot(1,3,3)
pylab.imshow(prediction_image,cmap='gray')
ax3.set_title('Prediction')
```

```
plt.show()
```





12 Experiments (5 P):

Compute The test set performance (loss on test test) for different noise values (1.5, 2.0, 2.5, 3.0, 3.5) and different regularizers C (0.1 0.5 0.9, 5, 10).

```
In [70]: noises = [1.5, 2., 2.5, 3., 3.5]
dataset = numpy.zeros((len(noises), 2))
for j, noise in enumerate(noises):
    x_test, y_test = make_toy_dataset(shape=shape, n_images=50, noise=noise)
    models_test = [build_model(x,y, weights) for x,y in zip(x_test, y_test)]
    losses = []
    for i, (gm, _, loss_function) in enumerate(models_test):
        lf = HammingLoss(y_test[i])

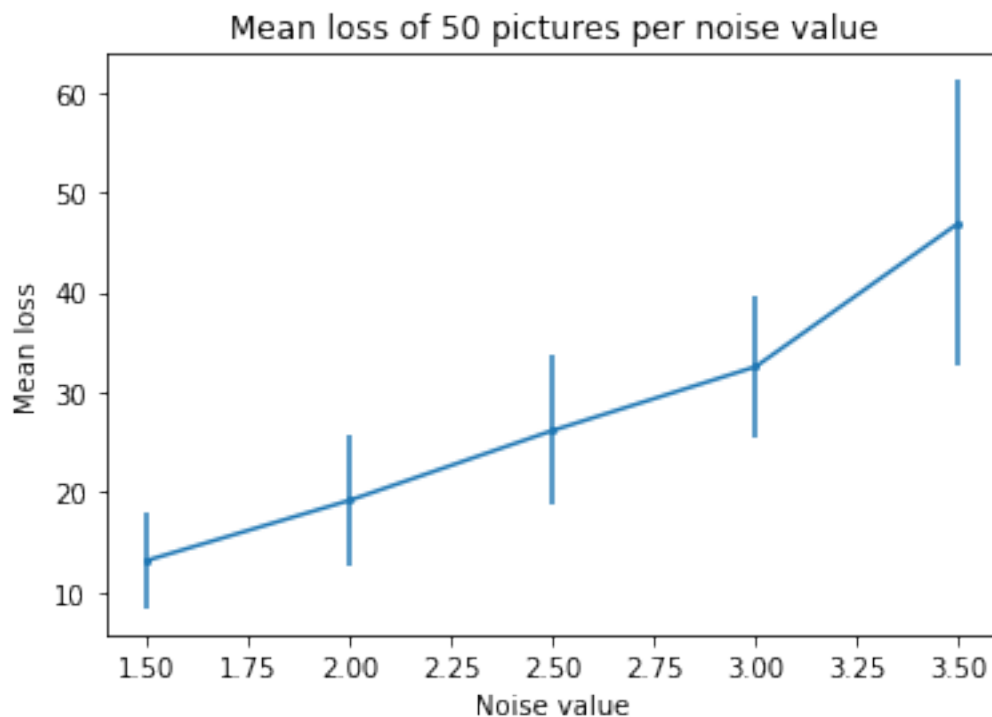
        gm.change_weights(weights)
        graphcut = GraphCut(model=gm)
        y_pred = graphcut.optimize()
        prediction_image = y_pred.reshape(shape)
```



```

        losses.append(lf(prediction_image))
    dataset[j, 0] = numpy.mean(losses)
    dataset[j, 1] = numpy.std(losses)
plt.cla()
plt.errorbar(noises, dataset[:, 0], marker='o', markersize=2, yerr=dataset[:, 1])
plt.title('Mean loss of 50 pictures per noise value')
plt.xlabel('Noise value')
plt.ylabel('Mean loss')
plt.show()

```



```

In [72]: Cs = [.1, .5, .9, 5, 10]
dataset = numpy.zeros((len(Cs), 2))
for j, C in enumerate(Cs):
    weights = subgradient_ssvm(dset,c=C,learning_rate=1.0, lower_bounds=lower_bounds,

    x_test , y_test = make_toy_dataset(shape=shape, n_images=50, noise=noise)
    models_test = [build_model(x,y, weights) for x,y in zip(x_test, y_test)]
    losses = []
    for i, (gm, _, loss_function) in enumerate(models_test):
        lf = HammingLoss(y_test[i])

        gm.change_weights(weights)
    graphcut = GraphCut(model=gm)

```

```

        y_pred = graphcut.optimize()
        prediction_image = y_pred.reshape(shape)

        losses.append(lf(prediction_image))
        dataset[j, 0] = numpy.mean(losses)
        dataset[j, 1] = numpy.std(losses)

iter 0 delta 29.6603896538    [ -6.41   -5.032  -1.588   2.23   14.4    0.     0.     0.
  0.   ]
iter 1 delta 63.1762670272    [-18.656 -16.907 -12.406  -7.108  -4.5     0.     0.     0.
  0.   ]
iter 2 delta 43.3172189859    [-12.629 -10.668  -5.724  -0.14   12.9     0.     0.     0.
  0.   ]
iter 3 delta 10.9121293799    [-11.773  -9.799  -4.833   0.734   14.86    0.769   0.977   0.975
  0.937   0.865]
iter 4 delta 16.8335039699    [-14.236 -12.168  -6.992  -1.163   10.936   0.164   0.375   0.346
  0.242   0.037]
iter 5 delta 23.0058643946    [-11.232  -9.059  -3.663   2.307   19.589    0.     0.     0.
  0.   ]
iter 6 delta 18.0503620078    [-14.731 -12.452  -6.754  -0.361   14.189    0.     0.     0.
  0.   ]
iter 7 delta 8.52923815112    [-13.988 -11.7    -5.979   0.41    15.964   0.504   0.672   0.675
  0.641   0.57 ]
iter 8 delta 1.50564230749    [-14.105 -11.808  -6.081   0.315   15.744   0.385   0.54    0.53
  0.484   0.412]
iter 9 delta 0.964085611698    [-14.151 -11.848  -6.117   0.28    15.658   0.286   0.428   0.408
  0.354   0.28 ]
iter 10 delta 0.83674041079    [-14.19   -11.882  -6.147   0.251   15.586   0.199   0.329   0.307
  0.241   0.164]
iter 11 delta 0.567399451723    [-14.201 -11.889  -6.153   0.246   15.566   0.127   0.247   0.217
  0.15    0.067]
iter 12 delta 0.476585850915    [-14.212 -11.896  -6.159   0.24    15.546   0.066   0.178   0.137
  0.072   0.   ]
iter 13 delta 0.295545932678    [-14.219 -11.901  -6.163   0.236   15.531   0.022   0.126   0.087
  0.016   0.   ]
iter 14 delta 0.167357767938    [-14.22   -11.899  -6.161   0.238   15.53    0.     0.088   0.047
  0.   ]
iter 15 delta 0.0452748587265    [-14.216 -11.894  -6.155   0.245   15.539    0.     0.081   0.037
  0.   ]
iter 16 delta 0.0292397413746    [-14.215 -11.891  -6.151   0.249   15.542    0.     0.075   0.033
  0.   ]
iter 17 delta 0.0276153112983    [-14.214 -11.888  -6.148   0.253   15.546    0.     0.069   0.031
  0.   ]
iter 18 delta 0.0146926642719    [ -1.42140000e+01  -1.18860000e+01  -6.14600000e+00   2.550000
  1.55470000e+01   1.00000000e-03   6.70000000e-02   1.60000000e-02
  0.00000000e+00   0.00000000e+00   0.00000000e+00]
iter 19 delta 0.0122530650533    [ -1.42150000e+01  -1.18860000e+01  -6.14500000e+00   2.570000
  1.55460000e+01   3.00000000e-03   6.50000000e-02   1.20000000e-02

```

```

0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 20 delta 0.011669585765 [-1.42160000e+01 -1.18850000e+01 -6.14400000e+00 2.580000
1.55450000e+01 4.00000000e-03 6.30000000e-02 9.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 21 delta 0.00949742466846 [-1.42170000e+01 -1.18850000e+01 -6.14300000e+00 2.5900
1.55430000e+01 8.00000000e-03 6.30000000e-02 8.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 22 delta 0.0089936066466 [-1.42180000e+01 -1.18850000e+01 -6.14200000e+00 2.60000
1.55420000e+01 1.10000000e-02 6.40000000e-02 8.00000000e-03
0.00000000e+00 1.00000000e-03 0.00000000e+00]
iter 23 delta 0.0107526020727 [-1.42180000e+01 -1.18840000e+01 -6.14100000e+00 2.62000
1.55420000e+01 1.20000000e-02 6.20000000e-02 5.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 24 delta 0.00827411811487 [-1.42190000e+01 -1.18840000e+01 -6.14000000e+00 2.6300
1.55410000e+01 1.50000000e-02 6.30000000e-02 4.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 25 delta 0.00968023849551 [-1.42200000e+01 -1.18830000e+01 -6.13900000e+00 2.6400
1.55400000e+01 1.60000000e-02 6.10000000e-02 1.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 26 delta 0.00766122047673 [-1.42200000e+01 -1.18830000e+01 -6.13900000e+00 2.6500
1.55400000e+01 1.90000000e-02 6.10000000e-02 1.00000000e-03
0.00000000e+00 0.00000000e+00 0.00000000e+00]
iter 27 delta 0.00738760545971 [-14.221 -11.883 -6.138 0.266 15.539 0.022 0.062 0
0. ]
iter 28 delta 0.00642775629849 [-14.221 -11.882 -6.137 0.267 15.539 0.023 0.06 0
0. ]
iter 29 delta 0.00568432708756 [-14.222 -11.882 -6.136 0.268 15.538 0.024 0.059 0
0. ]
iter 30 delta 0.00624280635919 [-14.223 -11.882 -6.136 0.269 15.537 0.026 0.059 0
0. ]
iter 31 delta 0.00556620492612 [-14.223 -11.881 -6.135 0.27 15.537 0.027 0.058 0
0. ]
iter 32 delta 0.0051675700796 [-14.224 -11.881 -6.134 0.271 15.536 0.028 0.056 0
0. ]
iter 33 delta 0.00501558272432 [-14.224 -11.88 -6.133 0.272 15.536 0.029 0.055 0
0. ]
iter 34 delta 0.00487228036077 [-14.225 -11.88 -6.133 0.273 15.535 0.03 0.054 0
0. ]
iter 35 delta 0.00473693923964 [-14.226 -11.88 -6.132 0.273 15.535 0.03 0.053 0
0. ]
iter 36 delta 0.00523045938203 [-14.226 -11.88 -6.132 0.274 15.534 0.032 0.053 0
0. ]
iter 37 delta 0.00567333128625 [-14.226 -11.879 -6.131 0.275 15.534 0.033 0.052 0
0. ]
iter 38 delta 0.00438917473592 [-14.227 -11.879 -6.131 0.276 15.533 0.034 0.051 0
0. ]
iter 39 delta 0.00426324531567 [-14.228 -11.879 -6.13 0.276 15.533 0.034 0.049 0
0. ]

```

iter 40	delta 0.00415926372261	[-14.228 -11.879 -6.13	0.277	15.532	0.035	0.048	0
0.]						
iter 41	delta 0.00460778564607	[-14.229 -11.879 -6.129	0.277	15.531	0.037	0.049	0
0.]						
iter 42	delta 0.00501299714133	[-14.229 -11.878 -6.128	0.279	15.532	0.037	0.048	0
0.]						
iter 43	delta 0.00389040487957	[-14.229 -11.878 -6.128	0.279	15.531	0.038	0.047	0
0.]						
iter 44	delta 0.00378955139171	[-14.23 -11.878 -6.128	0.28	15.53	0.039	0.046	0
0.]						
iter 45	delta 0.00370716983971	[-14.23 -11.877 -6.127	0.28	15.53	0.039	0.045	0
0.]						
iter 46	delta 0.00411759568372	[-14.231 -11.878 -6.127	0.281	15.529	0.041	0.045	0
0.]						
iter 47	delta 0.00449035544593	[-14.231 -11.877 -6.126	0.282	15.529	0.041	0.044	0
0.]						
iter 48	delta 0.00348020025769	[-14.231 -11.877 -6.126	0.282	15.529	0.042	0.043	0
0.]						
iter 49	delta 0.00341059625254	[-14.232 -11.876 -6.125	0.283	15.529	0.043	0.042	0
0.]						
iter 50	delta 0.00334372181621	[-14.232 -11.876 -6.125	0.283	15.528	0.043	0.041	0
0.]						
iter 51	delta 0.00372167302183	[-14.233 -11.876 -6.125	0.284	15.527	0.045	0.041	0
0.]						
iter 52	delta 0.00333625201309	[-14.233 -11.876 -6.124	0.284	15.527	0.045	0.041	0
0.]						
iter 53	delta 0.00315795949309	[-14.233 -11.876 -6.124	0.285	15.527	0.045	0.04	0
0.]						
iter 54	delta 0.00310054204776	[-14.233 -11.876 -6.123	0.286	15.527	0.046	0.039	0
0.]						
iter 55	delta 0.00304517522548	[-14.234 -11.875 -6.123	0.286	15.526	0.046	0.038	0
0.]						
iter 56	delta 0.00339521047605	[-14.234 -11.875 -6.123	0.286	15.526	0.048	0.038	0
0.]						
iter 57	delta 0.00371560193091	[-14.234 -11.875 -6.122	0.287	15.526	0.048	0.038	0
0.]						
iter 58	delta 0.00290131889324	[-14.235 -11.875 -6.122	0.288	15.525	0.049	0.037	0
0.]						
iter 59	delta 0.00284216354378	[-14.235 -11.875 -6.121	0.288	15.525	0.049	0.036	0
0.]						
iter 60	delta 0.00275480160479	[-14.235 -11.874 -6.121	0.289	15.525	0.05	0.035	0
0.]						
iter 61	delta 0.00329139513302	[-14.236 -11.875 -6.121	0.289	15.524	0.051	0.036	0
0.]						
iter 62	delta 0.00335627886686	[-14.236 -11.874 -6.12	0.29	15.524	0.051	0.035	0
0.]						
iter 63	delta 0.0026746533547	[-14.236 -11.874 -6.12	0.29	15.524	0.052	0.034	0.
0.]						

iter 64	delta 0.00258527535219	[-14.237 -11.874 -6.119	0.29	15.524	0.052	0.033	0
0.]						
iter 65	delta 0.00309191664011	[-14.237 -11.874 -6.119	0.291	15.523	0.053	0.034	0
0.]						
iter 66	delta 0.00315583645964	[-14.237 -11.874 -6.119	0.291	15.523	0.053	0.033	0
0.]						
iter 67	delta 0.00251732080443	[-14.238 -11.873 -6.119	0.292	15.523	0.054	0.032	0
0.]						
iter 68	delta 0.00243540431728	[-14.238 -11.873 -6.118	0.292	15.523	0.054	0.032	0
0.]						
iter 69	delta 0.00244539735287	[-14.238 -11.873 -6.118	0.292	15.522	0.054	0.031	0
0.]						
iter 70	delta 0.00268715988036	[-14.238 -11.873 -6.118	0.293	15.522	0.056	0.031	0
0.]						
iter 71	delta 0.00246423659582	[-14.239 -11.873 -6.117	0.293	15.521	0.056	0.03	0
0.]						
iter 72	delta 0.00230195750537	[-14.239 -11.873 -6.117	0.294	15.521	0.056	0.03	0
0.]						
iter 73	delta 0.0022796067563	[-14.239 -11.873 -6.117	0.294	15.521	0.056	0.029	0.
0.]						
iter 74	delta 0.00246245334026	[-14.239 -11.873 -6.117	0.294	15.521	0.057	0.029	0
0.]						
iter 75	delta 0.0022813492433	[-14.24 -11.872 -6.116	0.295	15.52	0.058	0.029	0.
0.]						
iter 76	delta 0.00219079090865	[-14.24 -11.872 -6.116	0.295	15.52	0.058	0.028	0
0.]						
iter 77	delta 0.00215439612682	[-14.24 -11.872 -6.116	0.296	15.52	0.058	0.028	0
0.]						
iter 78	delta 0.00233777215847	[-14.24 -11.872 -6.115	0.296	15.52	0.059	0.028	0
0.]						
iter 79	delta 0.00216724271016	[-14.241 -11.872 -6.115	0.296	15.519	0.06	0.027	0
0.]						
iter 80	delta 0.0020665306707	[-14.241 -11.872 -6.115	0.297	15.519	0.06	0.027	0.
0.]						
iter 81	delta 0.00205720609715	[-14.241 -11.872 -6.115	0.297	15.519	0.06	0.026	0
0.]						
iter 82	delta 0.00222510843999	[-14.241 -11.872 -6.114	0.297	15.519	0.061	0.026	0
0.]						
iter 83	delta 0.00204850805673	[-14.241 -11.871 -6.114	0.298	15.519	0.061	0.026	0
0.]						
iter 84	delta 0.00198459882313	[-14.242 -11.871 -6.114	0.298	15.518	0.061	0.025	0
0.]						
iter 85	delta 0.00196152209263	[-14.242 -11.871 -6.114	0.299	15.518	0.062	0.024	0
0.]						
iter 86	delta 0.00212280460367	[-14.242 -11.871 -6.113	0.299	15.518	0.063	0.025	0
0.]						
iter 87	delta 0.00195536490992	[-14.242 -11.871 -6.113	0.299	15.518	0.063	0.024	0
0.]						

iter 88	delta 0.00189540337041	[-14.243 -11.871 -6.113 0.3 15.517 0.063 0.024 0
0.]	
iter 89	delta 0.00187434333296	[-14.243 -11.871 -6.113 0.3 15.517 0.063 0.023 0
0.]	
iter 90	delta 0.0020294945112	[-14.243 -11.871 -6.112 0.3 15.517 0.064 0.023 0
0.]	
iter 91	delta 0.00187032361792	[-14.243 -11.87 -6.112 0.301 15.517 0.064 0.023 0
0.]	
iter 92	delta 0.0018138806448	[-14.243 -11.87 -6.112 0.301 15.517 0.065 0.022 0.
0.]	
iter 93	delta 0.00179458404219	[-14.244 -11.87 -6.112 0.301 15.516 0.065 0.022 0
0.]	
iter 94	delta 0.00194404211073	[-14.244 -11.87 -6.111 0.301 15.516 0.066 0.022 0
0.]	
iter 95	delta 0.0017923711522	[-14.244 -11.87 -6.111 0.302 15.516 0.066 0.021 0.
0.]	
iter 96	delta 0.00173908144295	[-14.244 -11.87 -6.111 0.302 15.516 0.066 0.021 0
0.]	
iter 97	delta 0.00172133571394	[-14.244 -11.87 -6.111 0.302 15.516 0.066 0.02 0
0.]	
iter 98	delta 0.00186549495474	[-14.245 -11.87 -6.111 0.303 15.515 0.067 0.02 0
0.]	
iter 99	delta 0.00172065661484	[-14.245 -11.87 -6.11 0.303 15.515 0.067 0.02 0
0.]	
iter 0	delta 148.301948269	[-32.05 -25.162 -7.94 11.149 72. 0. 0. 0.
0.]	
iter 1	delta 315.881335136	[-93.278 -84.535 -62.031 -35.541 -22.5 0. 0. 0.
0.]	
iter 2	delta 216.586094929	[-63.143 -53.341 -28.618 -0.698 64.5 0. 0. 0.
0.]	
iter 3	delta 15.1156490128	[-61.085 -51.222 -26.47 1.362 68.925 0.4 0.572 0.475
0.37	0.125]	
iter 4	delta 5.8495989785	[-61.669 -51.795 -27.047 0.812 67.665 0. 0. 0.
0.]	
iter 5	delta 0.770754366329	[-61.523 -51.65 -26.91 0.938 67.882 0. 0. 0.
0.]	
iter 6	delta 0.131381588897	[-61.531 -51.658 -26.923 0.923 67.796 0. 0. 0.
0.]	
iter 7	delta 0.069690615693	[-61.52 -51.647 -26.917 0.926 67.758 0. 0. 0.
0.]	
iter 8	delta 0.0619472139493	[-61.511 -51.636 -26.911 0.928 67.725 0. 0. 0.
0.]	
iter 9	delta 0.0835031517545	[-61.488 -51.612 -26.891 0.945 67.725 0. 0. 0.
0.]	
iter 10	delta 0.0506840841403	[-61.48 -51.603 -26.886 0.947 67.698 0. 0. 0.
0.]	
iter 11	delta 0.046460410462	[-61.473 -51.596 -26.881 0.949 67.673 0. 0. 0.
0.]	

iter 12	delta 0.0642331936573	[-61.455 -51.577 -26.866	0.961	67.673	0.	0.	0.
0.]						
iter 13	delta 0.0398232089674	[-61.449 -51.57 -26.862	0.963	67.651	0.	0.	0.
0.]						
iter 14	delta 0.0371683283696	[-61.443 -51.564 -26.859	0.965	67.631	0.	0.	0.
0.]						
iter 15	delta 0.0461726220169	[-61.432 -51.552 -26.849	0.972	67.625	0.	0.	0.
0.]						
iter 16	delta 0.0327955838555	[-61.427 -51.547 -26.846	0.974	67.608	0.	0.	0.
0.]						
iter 17	delta 0.0359898123466	[-61.419 -51.539 -26.841	0.977	67.596	0.	0.	0.
0.]						
iter 18	delta 0.0340956116968	[-61.412 -51.531 -26.835	0.981	67.586	0.	0.	0.
0.]						
iter 19	delta 0.0278762462772	[-61.408 -51.526 -26.832	0.982	67.571	0.	0.	0.
0.]						
iter 20	delta 0.0397634055974	[-61.397 -51.515 -26.823	0.99	67.571	0.	0.	0.
0.]						
iter 21	delta 0.0253420420702	[-61.393 -51.511 -26.821	0.991	67.557	0.	0.	0.
0.]						
iter 22	delta 0.0242402141541	[-61.39 -51.507 -26.818	0.992	67.544	0.	0.	0.
0.]						
iter 23	delta 0.0307817480113	[-61.382 -51.499 -26.812	0.997	67.54	0.	0.	0.
0.]						
iter 24	delta 0.0223009970218	[-61.379 -51.495 -26.81	0.998	67.528	0.	0.	0.
0.]						
iter 25	delta 0.0284139212412	[-61.371 -51.488 -26.804	1.002	67.524	0.	0.	0.
0.]						
iter 26	delta 0.0206490713164	[-61.368 -51.484 -26.802	1.003	67.513	0.	0.	0.
0.]						
iter 27	delta 0.0231363079371	[-61.364 -51.479 -26.798	1.006	67.506	0.	0.	0.
0.]						
iter 28	delta 0.0223385042151	[-61.359 -51.474 -26.795	1.008	67.499	0.	0.	0.
0.]						
iter 29	delta 0.0215938874079	[-61.355 -51.469 -26.791	1.011	67.492	0.	0.	0.
0.]						
iter 30	delta 0.0208973103948	[-61.35 -51.465 -26.788	1.013	67.486	0.	0.	0.
0.]						
iter 31	delta 0.0202442694449	[-61.346 -51.46 -26.785	1.015	67.48	0.	0.	0.
0.]						
iter 32	delta 0.0196308067345	[-61.342 -51.456 -26.782	1.017	67.474	0.	0.	0.
0.]						
iter 33	delta 0.0163977919278	[-61.34 -51.453 -26.78	1.018	67.465	0.	0.	0.
0.]						
iter 34	delta 0.0211074843506	[-61.335 -51.448 -26.776	1.021	67.462	0.	0.	0.
0.]						
iter 35	delta 0.0154868034873	[-61.332 -51.445 -26.774	1.022	67.454	0.	0.	0.
0.]						

iter 36	delta 0.0175085573578	[-61.329 -51.441 -26.772	1.024	67.448	0.	0.	0.
0.]						
iter 37	delta 0.0170478058484	[-61.325 -51.437 -26.769	1.026	67.443	0.	0.	0.
0.]						
iter 38	delta 0.0166106826215	[-61.322 -51.434 -26.766	1.027	67.438	0.	0.	0.
0.]						
iter 39	delta 0.016195415556	[-61.319 -51.43 -26.764	1.029	67.433	0.	0.	0.
0.]						
iter 40	delta 0.0158004054204	[-61.315 -51.427 -26.761	1.031	67.428	0.	0.	0.
0.]						
iter 41	delta 0.0154242052914	[-61.312 -51.423 -26.759	1.033	67.423	0.	0.	0.
0.]						
iter 42	delta 0.0150655028427	[-61.309 -51.42 -26.756	1.034	67.419	0.	0.	0.
0.]						
iter 43	delta 0.0147231050509	[-61.306 -51.417 -26.754	1.036	67.414	0.	0.	0.
0.]						
iter 44	delta 0.0143959249386	[-61.303 -51.414 -26.752	1.037	67.41	0.	0.	0.
0.]						
iter 45	delta 0.0140829700487	[-61.3 -51.41 -26.749	1.039	67.405	0.	0.	0.
0.]						
iter 46	delta 0.0137833323881	[-61.298 -51.407 -26.747	1.04	67.401	0.	0.	0.
0.]						
iter 47	delta 0.01349617963	[-61.295 -51.404 -26.745	1.042	67.397	0.	0.	0.
0.]						
iter 48	delta 0.011378059705	[-61.293 -51.402 -26.744	1.042	67.391	0.	0.	0.
0.]						
iter 49	delta 0.0129563324448	[-61.29 -51.4 -26.742	1.044	67.387	0.	0.	0.
0.]						
iter 50	delta 0.0127022867106	[-61.288 -51.397 -26.74	1.045	67.383	0.	0.	0.
0.]						
iter 51	delta 0.0124580119661	[-61.285 -51.394 -26.738	1.046	67.379	0.	0.	0.
0.]						
iter 52	delta 0.0122229551366	[-61.283 -51.391 -26.736	1.048	67.375	0.	0.	0.
0.]						
iter 53	delta 0.0119966041155	[-61.28 -51.389 -26.734	1.049	67.371	0.	0.	0.
0.]						
iter 54	delta 0.0117784840407	[-61.278 -51.386 -26.732	1.05	67.368	0.	0.	0.
0.]						
iter 55	delta 0.0115681539685	[-61.276 -51.384 -26.73	1.052	67.364	0.	0.	0.
0.]						
iter 56	delta 0.0113652038989	[-61.273 -51.381 -26.729	1.053	67.361	0.	0.	0.
0.]						
iter 57	delta 0.0111692521076	[-61.271 -51.379 -26.727	1.054	67.357	0.	0.	0.
0.]						
iter 58	delta 0.0109799427498	[-61.269 -51.376 -26.725	1.055	67.354	0.	0.	0.
0.]						
iter 59	delta 0.010796943704	[-61.267 -51.374 -26.723	1.056	67.35	0.	0.	0.
0.]						

iter 60	delta 0.0106199446269	[-61.264 -51.371 -26.722	1.058	67.347	0.	0.	0.
0.]						
iter 61	delta 0.0104486551974	[-61.262 -51.369 -26.72	1.059	67.344	0.	0.	0.
0.]						
iter 62	delta 0.0102828035276	[-61.26 -51.367 -26.718	1.06	67.341	0.	0.	0.
0.]						
iter 63	delta 0.0101221347225	[-61.258 -51.365 -26.717	1.061	67.338	0.	0.	0.
0.]						
iter 64	delta 0.00857730654683	[-61.257 -51.363 -26.716	1.061	67.333	0.	0.	0.
0.]						
iter 65	delta 0.00981540336725	[-61.255 -51.361 -26.714	1.062	67.33	0.	0.	0.
0.]						
iter 66	delta 0.00966890480953	[-61.253 -51.359 -26.713	1.063	67.327	0.	0.	0.
0.]						
iter 67	delta 0.00952671503291	[-61.251 -51.357 -26.711	1.064	67.324	0.	0.	0.
0.]						
iter 68	delta 0.00938864669911	[-61.249 -51.355 -26.71	1.065	67.321	0.	0.	0.
0.]						
iter 69	delta 0.00925452317483	[-61.247 -51.353 -26.708	1.066	67.318	0.	0.	0.
0.]						
iter 70	delta 0.00912417777801	[-61.245 -51.351 -26.707	1.067	67.316	0.	0.	0.
0.]						
iter 71	delta 0.00899745308665	[-61.244 -51.349 -26.705	1.068	67.313	0.	0.	0.
0.]						
iter 72	delta 0.00887420030464	[-61.242 -51.347 -26.704	1.069	67.31	0.	0.	0.
0.]						
iter 73	delta 0.00875427867891	[-61.24 -51.345 -26.703	1.07	67.307	0.	0.	0.
0.]						
iter 74	delta 0.00863755496318	[-61.238 -51.343 -26.701	1.071	67.305	0.	0.	0.
0.]						
iter 75	delta 0.0085239029242	[-61.236 -51.341 -26.7	1.072	67.302	0.	0.	0.
0.]						
iter 76	delta 0.00841320288621	[-61.235 -51.339 -26.698	1.073	67.299	0.	0.	0.
0.]						
iter 77	delta 0.00830534131075	[-61.233 -51.337 -26.697	1.074	67.297	0.	0.	0.
0.]						
iter 78	delta 0.00820021040808	[-61.231 -51.335 -26.696	1.075	67.294	0.	0.	0.
0.]						
iter 79	delta 0.0069690615693	[-61.23 -51.334 -26.695	1.075	67.291	0.	0.	0.
0.]						
iter 80	delta 0.00799773607702	[-61.229 -51.333 -26.694	1.076	67.288	0.	0.	0.
0.]						
iter 81	delta 0.00790020271022	[-61.227 -51.331 -26.693	1.077	67.286	0.	0.	0.
0.]						
iter 82	delta 0.00780501954504	[-61.226 -51.329 -26.691	1.078	67.283	0.	0.	0.
0.]						
iter 83	delta 0.00771210264569	[-61.224 -51.327 -26.69	1.079	67.281	0.	0.	0.
0.]						

iter 84	delta 0.00762137202633	[-61.222 -51.326 -26.689	1.079	67.279	0.	0.	0
0.]						
iter 85	delta 0.00753275142138	[-61.221 -51.324 -26.688	1.08	67.276	0.	0.	0
0.]						
iter 86	delta 0.00744616807172	[-61.219 -51.322 -26.687	1.081	67.274	0.	0.	0
0.]						
iter 87	delta 0.00736155252543	[-61.218 -51.321 -26.685	1.082	67.272	0.	0.	0
0.]						
iter 88	delta 0.00727883845212	[-61.216 -51.319 -26.684	1.083	67.269	0.	0.	0
0.]						
iter 89	delta 0.00719796246931	[-61.215 -51.318 -26.683	1.083	67.267	0.	0.	0
0.]						
iter 90	delta 0.00711886398064	[-61.213 -51.316 -26.682	1.084	67.265	0.	0.	0
0.]						
iter 91	delta 0.00704148502433	[-61.212 -51.314 -26.681	1.085	67.263	0.	0.	0
0.]						
iter 92	delta 0.00696577013159	[-61.211 -51.313 -26.68	1.086	67.261	0.	0.	0
0.]						
iter 93	delta 0.00593111622918	[-61.21 -51.312 -26.679	1.086	67.257	0.	0.	0
0.]						
iter 94	delta 0.00681912233936	[-61.208 -51.31 -26.678	1.087	67.255	0.	0.	0
0.]						
iter 95	delta 0.00674808981498	[-61.207 -51.309 -26.677	1.087	67.253	0.	0.	0
0.]						
iter 96	delta 0.00667852187875	[-61.206 -51.307 -26.676	1.088	67.251	0.	0.	0
0.]						
iter 97	delta 0.00661037369631	[-61.204 -51.306 -26.675	1.089	67.249	0.	0.	0
0.]						
iter 98	delta 0.00654360224483	[-61.203 -51.305 -26.674	1.089	67.247	0.	0.	0
0.]						
iter 99	delta 0.00647816622238	[-61.202 -51.303 -26.673	1.09	67.245	0.	0.	0
0.]						
iter 0	delta 266.943506884	[-57.691 -45.292 -14.293	20.068	129.6	0.	0.	
0.	0.]						
iter 1	delta 568.586403245	[-167.9 -152.163 -111.656	-63.974	-40.5	0.	0.	
0.	0.]						
iter 2	delta 389.854970873	[-113.657 -96.013 -51.512	-1.257	116.1	0.	0.	
0.	0.]						
iter 3	delta 24.6526479912	[-110.105 -92.357 -47.812	2.292	123.705	0.495	0.739	
0.388	0.392 0.]						
iter 4	delta 7.5790676712	[-110.902 -93.151 -48.627	1.509	121.905	0.	0.	
0.	0.]						
iter 5	delta 0.969161533781	[-110.714 -92.958 -48.445	1.675	122.145	0.	0.	
0.	0.]						
iter 6	delta 0.167681809831	[-110.709 -92.958 -48.459	1.655	122.016	0.	0.	
0.	0.]						
iter 7	delta 0.146721583602	[-110.705 -92.957 -48.472	1.638	121.904	0.	0.	
0.	0.]						

iter 8	delta 0.115294996725	[-110.683 -92.937 -48.463	1.641 121.844	0.	0.
	0. 0.]				
iter 9	delta 0.103765497052	[-110.663 -92.918 -48.455	1.644 121.79	0.	0.
	0. 0.]				
iter 10	delta 0.109811240516	[-110.637 -92.893 -48.439	1.654 121.757	0.	0.
	0. 0.]				
iter 11	delta 0.100660303806	[-110.614 -92.869 -48.424	1.664 121.727	0.	0.
	0. 0.]				
iter 12	delta 0.0929172035134	[-110.592 -92.847 -48.411	1.672 121.7	0.	0.
	0. 0.]				
iter 13	delta 0.0741182121802	[-110.578 -92.834 -48.405	1.674 121.661	0.	0.
	0. 0.]				
iter 14	delta 0.080528243045	[-110.559 -92.815 -48.393	1.682 121.637	0.	0.
	0. 0.]				
iter 15	delta 0.0754952278546	[-110.541 -92.798 -48.383	1.689 121.614	0.	0.
	0. 0.]				
iter 16	delta 0.0710543320985	[-110.525 -92.781 -48.372	1.695 121.593	0.	0.
	0. 0.]				
iter 17	delta 0.0671068692041	[-110.509 -92.766 -48.363	1.701 121.573	0.	0.
	0. 0.]				
iter 18	delta 0.0635749287197	[-110.494 -92.751 -48.354	1.707 121.554	0.	0.
	0. 0.]				
iter 19	delta 0.0603961822837	[-110.48 -92.737 -48.345	1.713 121.536	0.	0.
	0. 0.]				
iter 20	delta 0.0494121414535	[-110.471 -92.728 -48.341	1.714 121.511	0.	0.
	0. 0.]				
iter 21	delta 0.0549056202579	[-110.458 -92.715 -48.333	1.719 121.494	0.	0.
	0. 0.]				
iter 22	delta 0.0525184193771	[-110.446 -92.703 -48.325	1.724 121.479	0.	0.
	0. 0.]				
iter 23	delta 0.0503301519031	[-110.434 -92.691 -48.318	1.729 121.464	0.	0.
	0. 0.]				
iter 24	delta 0.048316945827	[-110.423 -92.68 -48.311	1.733 121.449	0.	0.
	0. 0.]				
iter 25	delta 0.0464586017567	[-110.412 -92.669 -48.305	1.737 121.435	0.	0.
	0. 0.]				
iter 26	delta 0.0447379128028	[-110.401 -92.659 -48.298	1.741 121.422	0.	0.
	0. 0.]				
iter 27	delta 0.0431401302027	[-110.391 -92.649 -48.292	1.745 121.409	0.	0.
	0. 0.]				
iter 28	delta 0.0357812058801	[-110.384 -92.642 -48.289	1.746 121.391	0.	0.
	0. 0.]				
iter 29	delta 0.0402641215225	[-110.375 -92.633 -48.283	1.75 121.379	0.	0.
	0. 0.]				
iter 30	delta 0.0389652788927	[-110.366 -92.624 -48.278	1.754 121.367	0.	0.
	0. 0.]				
iter 31	delta 0.0377476139273	[-110.357 -92.615 -48.272	1.757 121.356	0.	0.
	0. 0.]				

iter 32	delta 0.0366037468386	[-110.349	-92.606	-48.267	1.761	121.345	0.	0.
	0. 0.]							
iter 33	delta 0.0355271660493	[-110.34	-92.598	-48.262	1.764	121.334	0.	0.
	0. 0.]							
iter 34	delta 0.0345121041621	[-110.332	-92.59	-48.257	1.767	121.324	0.	0.
	0. 0.]							
iter 35	delta 0.0288237491812	[-110.327	-92.585	-48.255	1.768	121.309	0.	0.
	0. 0.]							
iter 36	delta 0.0326465850182	[-110.319	-92.577	-48.25	1.771	121.299	0.	0.
	0. 0.]							
iter 37	delta 0.0317874643599	[-110.312	-92.57	-48.245	1.774	121.29	0.	0.
	0. 0.]							
iter 38	delta 0.0309724011712	[-110.305	-92.563	-48.241	1.777	121.28	0.	0.
	0. 0.]							
iter 39	delta 0.0301980911419	[-110.298	-92.556	-48.237	1.779	121.271	0.	0.
	0. 0.]							
iter 40	delta 0.0294615523335	[-110.291	-92.549	-48.232	1.782	121.263	0.	0.
	0. 0.]							
iter 41	delta 0.0247060707267	[-110.286	-92.544	-48.23	1.783	121.25	0.	0.
	0. 0.]							
iter 42	delta 0.0280912475738	[-110.279	-92.538	-48.226	1.785	121.241	0.	0.
	0. 0.]							
iter 43	delta 0.027452810129	[-110.273	-92.531	-48.222	1.788	121.233	0.	0.
	0. 0.]							
iter 44	delta 0.0268427476816	[-110.267	-92.525	-48.219	1.79	121.225	0.	0.
	0. 0.]							
iter 45	delta 0.0262592096886	[-110.261	-92.519	-48.215	1.793	121.217	0.	0.
	0. 0.]							
iter 46	delta 0.0257005030995	[-110.255	-92.513	-48.211	1.795	121.21	0.	0.
	0. 0.]							
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iter 3 delta 234.624743165    [ -1.22658700e+03  -1.02955100e+03  -5.34653000e+02  2.21590000e+00
1.36700000e+03  2.47300000e+00  4.13700000e+00  2.41600000e+00
5.38000000e-01  7.01000000e-01  0.00000000e+00]
iter 4 delta 46.7956913275    [-1232.215 -1035.21   -540.547    16.409  1353.4      0.
0.      0.      0.      ]
iter 5 delta 7.8211173194     [-1230.584 -1033.532  -539.043    17.75   1355.067    0.
0.      0.      0.      ]
iter 6 delta 2.50616669892     [-1229.865 -1032.797  -538.46     18.22   1355.067    0.
0.      0.      0.      ]
iter 7 delta 1.71543899566     [-1229.471 -1032.403  -538.203    18.39   1354.567    0.
0.      0.      0.      ]
iter 8 delta 1.52483466281     [-1229.121 -1032.052  -537.975    18.541  1354.122    0.
0.      0.      0.      ]
iter 9 delta 1.37235119652     [-1228.806 -1031.736  -537.77     18.677  1353.722    0.
0.      0.      0.      ]
iter 10 delta 1.24759199684     [-1228.52  -1031.449  -537.583    18.801  1353.359    0.
0.      0.      0.      ]
iter 11 delta 1.1436259971     [-1228.257 -1031.186  -537.412    18.914  1353.025    0.
0.      0.      0.      ]
iter 12 delta 1.05565476656     [-1228.015 -1030.943  -537.254    19.019  1352.718    0.
0.      0.      0.      ]
iter 13 delta 0.980250854661     [-1227.79  -1030.717  -537.107    19.116  1352.432    0.
0.      0.      0.      ]
iter 14 delta 0.914900797683     [-1227.58  -1030.506  -536.97     19.207  1352.165    0.
0.      0.      0.      ]
iter 15 delta 0.857719497828     [-1227.383 -1030.309  -536.842    19.292  1351.915    0.
0.      0.      0.      ]
iter 16 delta 0.807265409721     [-1227.197 -1030.123  -536.721    19.372  1351.68     0.
0.      0.      0.      ]
iter 17 delta 0.762417331403     [-1227.022 -1029.948  -536.607    19.447  1351.458    0.
0.      0.      0.      ]
iter 18 delta 0.722290103434     [-1226.856 -1029.782  -536.499    19.519  1351.247    0.
0.      0.      0.      ]
iter 19 delta 0.686175598262     [-1226.699 -1029.624  -536.396    19.587  1351.047    0.
0.      0.      0.      ]
iter 20 delta 0.653500569774     [-1226.549 -1029.473  -536.299    19.652  1350.857    0.
0.      0.      0.      ]
iter 21 delta 0.62379599842     [-1226.406 -1029.33   -536.205    19.714  1350.675    0.
0.      0.      0.      ]
iter 22 delta 0.596674433272     [-1226.269 -1029.192  -536.116    19.773  1350.501    0.
0.      0.      0.      ]

```

iter 23	delta 0.571812998552	[-1226.137 -1029.061 -536.03	19.829	1350.334	0.
0.	0. 0.]				
iter 24	delta 0.54894047861	[-1226.011 -1028.935 -535.948	19.884	1350.174	0.
0.	0. 0.]				
iter 25	delta 0.527827383279	[-1225.89 -1028.813 -535.869	19.936	1350.02	0.
0.	0. 0.]				
iter 26	delta 0.508278220935	[-1225.773 -1028.696 -535.793	19.987	1349.872	0.
0.	0. 0.]				
iter 27	delta 0.49012542733	[-1225.661 -1028.583 -535.72	20.035	1349.729	0.
0.	0. 0.]				
iter 28	delta 0.473224550526	[-1225.552 -1028.474 -535.649	20.082	1349.591	0.
0.	0. 0.]				
iter 29	delta 0.457450398842	[-1225.447 -1028.369 -535.581	20.127	1349.458	0.
0.	0. 0.]				
iter 30	delta 0.442693934363	[-1225.345 -1028.267 -535.514	20.171	1349.329	0.
0.	0. 0.]				
iter 31	delta 0.428859748914	[-1225.247 -1028.168 -535.45	20.214	1349.204	0.
0.	0. 0.]				
iter 32	delta 0.415863998947	[-1225.151 -1028.073 -535.388	20.255	1349.083	0.
0.	0. 0.]				
iter 33	delta 0.40363270486	[-1225.059 -1027.98 -535.327	20.295	1348.965	0.
0.	0. 0.]				
iter 34	delta 0.392100341864	[-1224.969 -1027.89 -535.269	20.334	1348.851	0.
0.	0. 0.]				
iter 35	delta 0.381208665701	[-1224.881 -1027.802 -535.212	20.372	1348.74	0.
0.	0. 0.]				
iter 36	delta 0.370905728791	[-1224.796 -1027.717 -535.156	20.408	1348.632	0.
0.	0. 0.]				
iter 37	delta 0.361145051717	[-1224.713 -1027.633 -535.102	20.444	1348.526	0.
0.	0. 0.]				
iter 38	delta 0.351884922186	[-1224.632 -1027.552 -535.05	20.479	1348.424	0.
0.	0. 0.]				
iter 39	delta 0.343087799131	[-1224.554 -1027.474 -534.998	20.513	1348.324	0.
0.	0. 0.]				
iter 40	delta 0.33471980403	[-1224.477 -1027.396 -534.948	20.546	1348.226	0.
0.	0. 0.]				
iter 41	delta 0.326750284887	[-1224.402 -1027.321 -534.899	20.579	1348.131	0.
0.	0. 0.]				
iter 42	delta 0.319151441052	[-1224.328 -1027.248 -534.851	20.61	1348.038	0.
0.	0. 0.]				
iter 43	delta 0.31189799921	[-1224.257 -1027.176 -534.805	20.641	1347.947	0.
0.	0. 0.]				
iter 44	delta 0.304966932561	[-1224.187 -1027.106 -534.759	20.672	1347.858	0.
0.	0. 0.]				
iter 45	delta 0.298337216636	[-1224.118 -1027.037 -534.714	20.701	1347.771	0.
0.	0. 0.]				
iter 46	delta 0.291989616282	[-1224.051 -1026.97 -534.671	20.73	1347.686	0.
0.	0. 0.]				

iter 47	delta 0.285906499276	[-1223.986 -1026.904 -534.628	20.758	1347.603	0.
0.	0. 0.]				
iter 48	delta 0.28007167276	[-1223.921 -1026.84 -534.586	20.786	1347.521	0.
0.	0. 0.]				
iter 49	delta 0.274470239305	[-1223.858 -1026.777 -534.545	20.813	1347.441	0.
0.	0. 0.]				
iter 50	delta 0.269088469907	[-1223.796 -1026.715 -534.505	20.84	1347.363	0.
0.	0. 0.]				
iter 51	delta 0.263913691639	[-1223.736 -1026.654 -534.465	20.866	1347.286	0.
0.	0. 0.]				
iter 52	delta 0.258934188024	[-1223.676 -1026.594 -534.427	20.892	1347.21	0.
0.	0. 0.]				
iter 53	delta 0.254139110468	[-1223.618 -1026.536 -534.388	20.917	1347.136	0.
0.	0. 0.]				
iter 54	delta 0.249518399368	[-1223.561 -1026.478 -534.351	20.942	1347.064	0.
0.	0. 0.]				
iter 55	delta 0.245062713665	[-1223.504 -1026.422 -534.314	20.966	1346.992	0.
0.	0. 0.]				
iter 56	delta 0.240763367811	[-1223.449 -1026.367 -534.278	20.99	1346.922	0.
0.	0. 0.]				
iter 57	delta 0.236612275263	[-1223.395 -1026.312 -534.243	21.013	1346.853	0.
0.	0. 0.]				
iter 58	delta 0.232601897716	[-1223.341 -1026.259 -534.208	21.036	1346.785	0.
0.	0. 0.]				
iter 59	delta 0.228725199421	[-1223.289 -1026.206 -534.174	21.059	1346.719	0.
0.	0. 0.]				
iter 60	delta 0.224975605987	[-1223.237 -1026.154 -534.14	21.081	1346.653	0.
0.	0. 0.]				
iter 61	delta 0.221346967181	[-1223.186 -1026.103 -534.107	21.103	1346.589	0.
0.	0. 0.]				
iter 62	delta 0.217833523258	[-1223.136 -1026.053 -534.075	21.125	1346.525	0.
0.	0. 0.]				
iter 63	delta 0.214429874457	[-1223.087 -1026.004 -534.043	21.146	1346.463	0.
0.	0. 0.]				
iter 64	delta 0.211130953312	[-1223.039 -1025.955 -534.011	21.167	1346.401	0.
0.	0. 0.]				
iter 65	delta 0.207931999473	[-1222.991 -1025.907 -533.98	21.188	1346.34	0.
0.	0. 0.]				
iter 66	delta 0.204828536795	[-1222.944 -1025.86 -533.949	21.208	1346.281	0.
0.	0. 0.]				
iter 67	delta 0.20181635243	[-1222.898 -1025.814 -533.919	21.228	1346.222	0.
0.	0. 0.]				
iter 68	delta 0.198891477757	[-1222.852 -1025.768 -533.889	21.248	1346.164	0.
0.	0. 0.]				
iter 69	delta 0.196050170932	[-1222.807 -1025.723 -533.86	21.267	1346.107	0.
0.	0. 0.]				
iter 70	delta 0.193288900919	[-1222.763 -1025.678 -533.831	21.286	1346.05	0.
0.	0. 0.]				

iter 71	delta 0.190604332851	[-1222.719 -1025.635 -533.802	21.305	1345.995	0.
0.	0. 0.]				
iter 72	delta 0.187993314593	[-1222.676 -1025.591 -533.774	21.324	1345.94	0.
0.	0. 0.]				
iter 73	delta 0.185452864395	[-1222.633 -1025.549 -533.747	21.342	1345.886	0.
0.	0. 0.]				
iter 74	delta 0.182980159537	[-1222.591 -1025.506 -533.719	21.361	1345.833	0.
0.	0. 0.]				
iter 75	delta 0.180572525858	[-1222.55 -1025.465 -533.692	21.378	1345.78	0.
0.	0. 0.]				
iter 76	delta 0.17822742812	[-1222.509 -1025.424 -533.665	21.396	1345.728	0.
0.	0. 0.]				
iter 77	delta 0.175942461093	[-1222.468 -1025.383 -533.639	21.414	1345.677	0.
0.	0. 0.]				
iter 78	delta 0.173715341332	[-1222.428 -1025.343 -533.613	21.431	1345.626	0.
0.	0. 0.]				
iter 79	delta 0.171543899565	[-1222.389 -1025.304 -533.587	21.448	1345.576	0.
0.	0. 0.]				
iter 80	delta 0.169426073645	[-1222.35 -1025.265 -533.562	21.465	1345.527	0.
0.	0. 0.]				
iter 81	delta 0.167359902015	[-1222.312 -1025.226 -533.537	21.481	1345.478	0.
0.	0. 0.]				
iter 82	delta 0.165343517654	[-1222.274 -1025.188 -533.512	21.498	1345.43	0.
0.	0. 0.]				
iter 83	delta 0.163375142443	[-1222.236 -1025.151 -533.488	21.514	1345.382	0.
0.	0. 0.]				
iter 84	delta 0.161453081944	[-1222.199 -1025.114 -533.464	21.53	1345.335	0.
0.	0. 0.]				
iter 85	delta 0.159575720526	[-1222.162 -1025.077 -533.44	21.546	1345.289	0.
0.	0. 0.]				
iter 86	delta 0.157741516842	[-1222.126 -1025.041 -533.416	21.561	1345.243	0.
0.	0. 0.]				
iter 87	delta 0.155948999605	[-1222.09 -1025.005 -533.393	21.577	1345.197	0.
0.	0. 0.]				
iter 88	delta 0.154196763655	[-1222.055 -1024.969 -533.37	21.592	1345.152	0.
0.	0. 0.]				
iter 89	delta 0.152483466281	[-1222.02 -1024.934 -533.347	21.607	1345.108	0.
0.	0. 0.]				
iter 90	delta 0.150807823794	[-1221.985 -1024.899 -533.324	21.622	1345.064	0.
0.	0. 0.]				
iter 91	delta 0.149168608318	[-1221.951 -1024.865 -533.302	21.637	1345.02	0.
0.	0. 0.]				
iter 92	delta 0.147564644787	[-1221.917 -1024.831 -533.28	21.651	1344.977	0.
0.	0. 0.]				
iter 93	delta 0.145994808141	[-1221.884 -1024.798 -533.258	21.666	1344.935	0.
0.	0. 0.]				
iter 94	delta 0.144458020687	[-1221.851 -1024.764 -533.237	21.68	1344.893	0.
0.	0. 0.]				

```

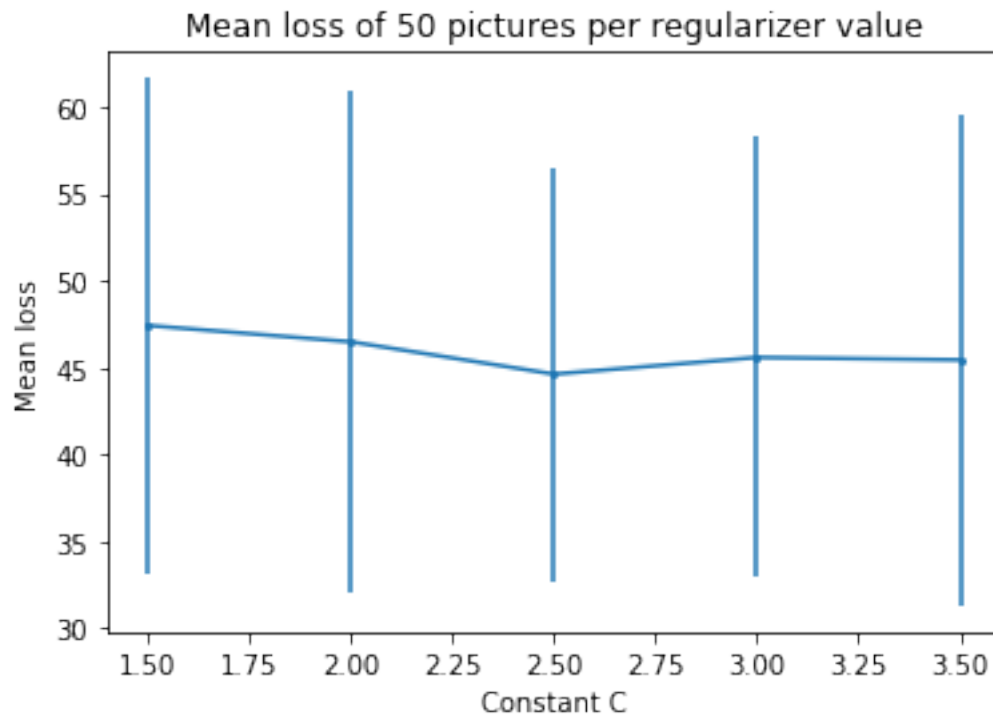
iter 95 delta 0.142953249638 [-1221.818 -1024.731 -533.215 21.694 1344.851 0.
0. 0. 0. ]
iter 96 delta 0.141479504796 [-1221.785 -1024.699 -533.194 21.708 1344.81 0.
0. 0. 0. ]
iter 97 delta 0.14003583638 [-1221.753 -1024.667 -533.173 21.722 1344.769 0.
0. 0. 0. ]
iter 98 delta 0.138621332982 [-1221.721 -1024.635 -533.152 21.736 1344.729 0.
0. 0. 0. ]
iter 99 delta 0.137235119652 [-1221.69 -1024.603 -533.132 21.75 1344.689 0.
0. 0. 0. ]

```

```

In [73]: plt.cla()
plt.errorbar(noises, dataset[:, 0], marker='o', markersize=2, yerr=dataset[:, 1])
plt.title('Mean loss of 50 pictures per regularizer value')
plt.xlabel('Constant C')
plt.ylabel('Mean loss')
plt.show()

```



13 Experiments (5 P):

Replace GraphCut with IteratedConditionalModes (ICM) to see if optimality is important.

```

In [74]: lower_bounds = numpy.ones(len(weights))*(-1.0*float('inf'))

# we want the regularizer 'beta' to be positive
lower_bounds[n_unary_features:n_unary_features+n_potts_features] = 0

weights = subgradient_ssvm(dset,c=0.5,learning_rate=1.0, lower_bounds=lower_bounds, n
                           algorithm=IteratedConditionalModes)

for i,(gm,_,loss_function) in enumerate(models_test):
    gm.change_weights(weights)

    a = IteratedConditionalModes(model=gm)
    y_pred = a.optimize()

    prediction_image = y_pred.reshape(shape)

    # show a bit from the dataset

    f= pylab.figure()
    ax1 = f.add_subplot(1,3,1)
    pylab.imshow(x_test[i],cmap='gray')
    ax1.set_title('Raw Data')

    ax2 = f.add_subplot(1,3,2)
    pylab.imshow(y_test[i],cmap='gray')
    ax2.set_title('Ground Truth')

    ax3 = f.add_subplot(1,3,3)
    pylab.imshow(prediction_image,cmap='gray')
    ax3.set_title('Prediction')

    plt.show()

iter 0 delta 148.301948269    [-32.05 -25.162 -7.94  11.149  72.      0.      0.      0.
0.    ]
iter 1 delta 315.881335136    [-93.278 -84.535 -62.031 -35.541 -22.5      0.      0.      0.
0.    ]
iter 2 delta 216.586094929    [-63.143 -53.341 -28.618 -0.698  64.5      0.      0.      0.
0.    ]
iter 3 delta 15.1156490128    [-61.085 -51.222 -26.47   1.362  68.925  0.4      0.572  0.475
0.37   0.125]
iter 4 delta 6.02735133396    [-61.701 -51.825 -27.075   0.785  67.605  0.      0.      0.
0.    ]
iter 5 delta 0.869048533361    [-61.54  -51.663 -26.921   0.927  67.855  0.      0.      0.
0.    ]
iter 6 delta 0.0723001606803    [-61.535 -51.657 -26.921   0.924  67.798  0.      0.      0.
0.    ]

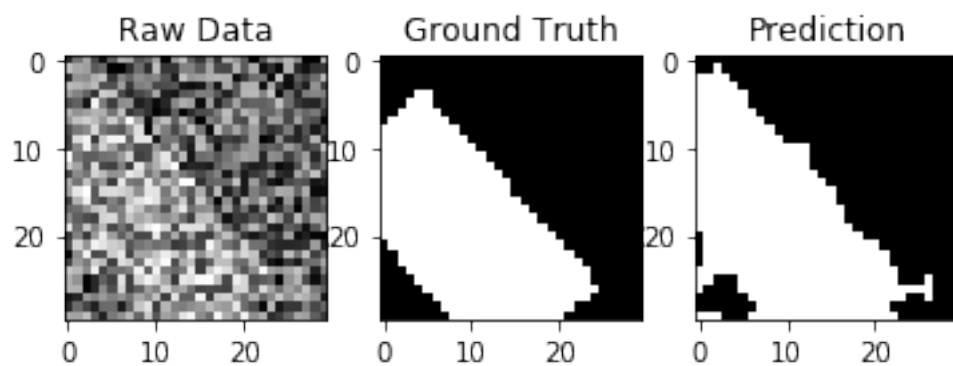
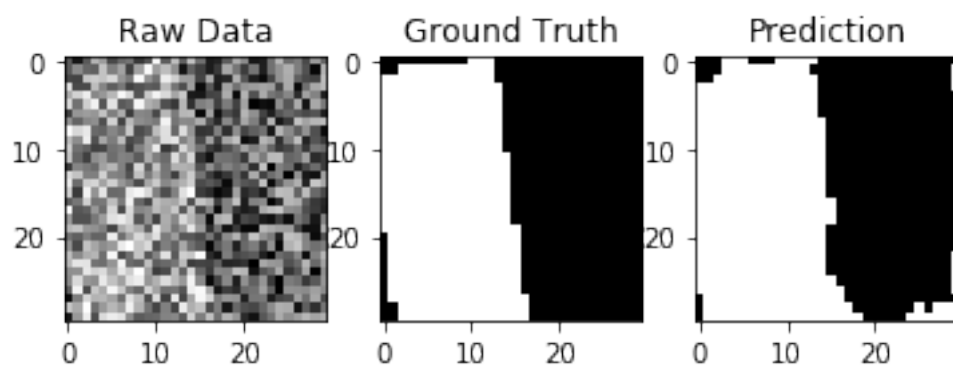
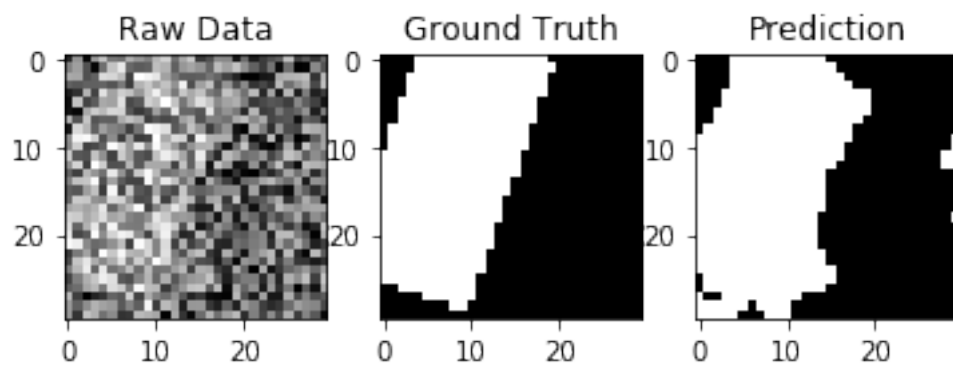
```

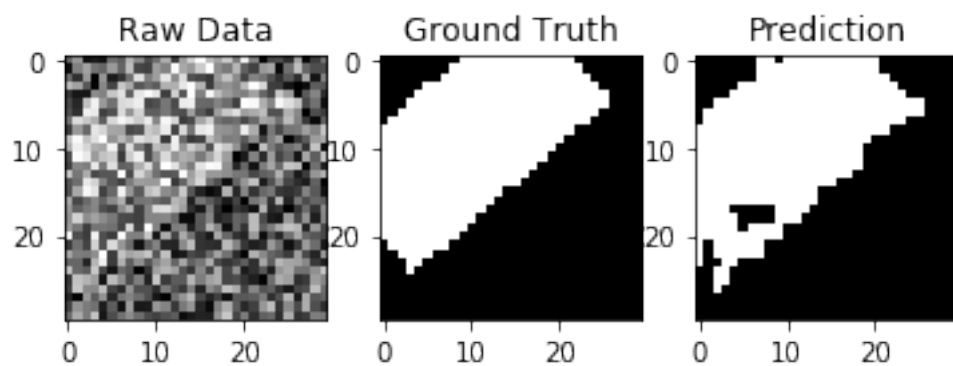
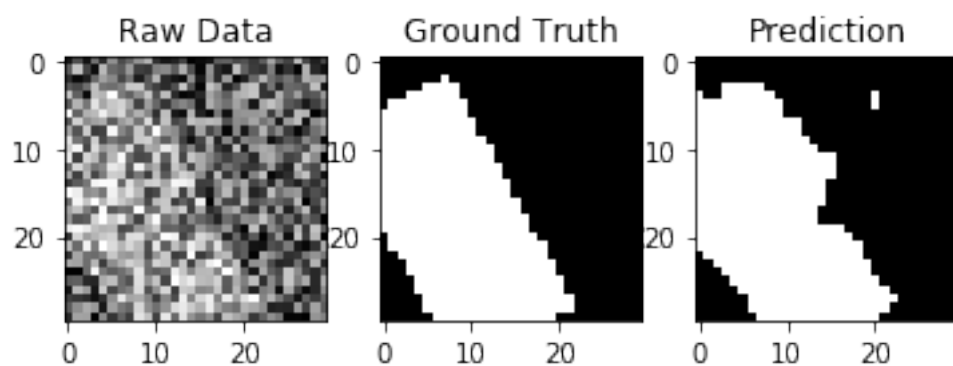
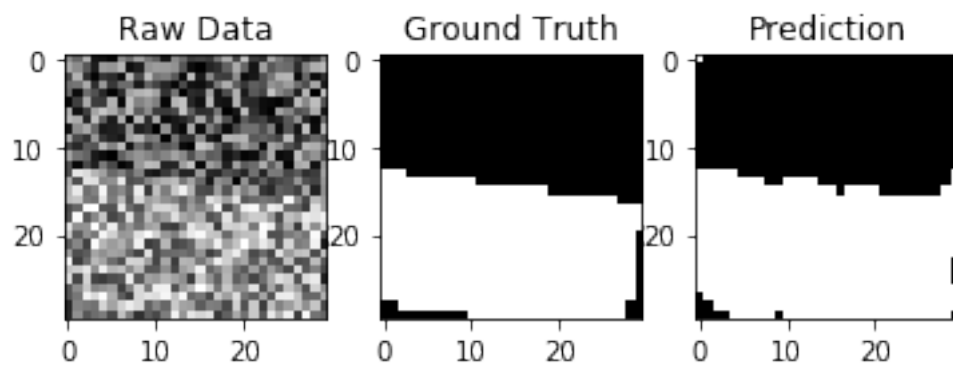

iter 7	delta 0.069690615693	[-61.524 -51.645 -26.914	0.927	67.76	0.	0.	0.
0.]						
iter 8	delta 0.0619472139493	[-61.515 -51.635 -26.908	0.93	67.727	0.	0.	0.
0.]						
iter 9	delta 0.0835031517545	[-61.492 -51.611 -26.888	0.946	67.727	0.	0.	0.
0.]						
iter 10	delta 0.0506840841403	[-61.484 -51.602 -26.883	0.949	67.7	0.	0.	0.
0.]						
iter 11	delta 0.046460410462	[-61.477 -51.594 -26.879	0.951	67.675	0.	0.	0.
0.]						
iter 12	delta 0.0642331936573	[-61.459 -51.576 -26.864	0.963	67.675	0.	0.	0.
0.]						
iter 13	delta 0.0398232089674	[-61.453 -51.569 -26.86	0.965	67.653	0.	0.	0.
0.]						
iter 14	delta 0.0371683283696	[-61.447 -51.563 -26.856	0.967	67.633	0.	0.	0.
0.]						
iter 15	delta 0.0404885388899	[-61.439 -51.554 -26.85	0.971	67.621	0.	0.	0.
0.]						
iter 16	delta 0.0327955838555	[-61.434 -51.548 -26.847	0.972	67.603	0.	0.	0.
0.]						
iter 17	delta 0.0463906398636	[-61.421 -51.535 -26.836	0.981	67.603	0.	0.	0.
0.]						
iter 18	delta 0.0293434171339	[-61.416 -51.53 -26.833	0.983	67.587	0.	0.	0.
0.]						
iter 19	delta 0.0278762462772	[-61.412 -51.525 -26.83	0.984	67.572	0.	0.	0.
0.]						
iter 20	delta 0.0351791405843	[-61.403 -51.516 -26.823	0.989	67.568	0.	0.	0.
0.]						
iter 21	delta 0.0253420420702	[-61.399 -51.512 -26.821	0.991	67.554	0.	0.	0.
0.]						
iter 22	delta 0.0321200848813	[-61.391 -51.504 -26.814	0.996	67.55	0.	0.	0.
0.]						
iter 23	delta 0.023230205231	[-61.388 -51.5 -26.812	0.997	67.537	0.	0.	0.
0.]						
iter 24	delta 0.0259126648895	[-61.383 -51.494 -26.808	0.999	67.529	0.	0.	0.
0.]						
iter 25	delta 0.0249160239322	[-61.378 -51.488 -26.804	1.002	67.521	0.	0.	0.
0.]						
iter 26	delta 0.0239932082311	[-61.373 -51.483 -26.8	1.005	67.514	0.	0.	0.
0.]						
iter 27	delta 0.0231363079371	[-61.368 -51.478 -26.796	1.007	67.507	0.	0.	0.
0.]						
iter 28	delta 0.0223385042151	[-61.363 -51.473 -26.793	1.01	67.5	0.	0.	0.
0.]						
iter 29	delta 0.0215938874079	[-61.359 -51.468 -26.789	1.012	67.493	0.	0.	0.
0.]						
iter 30	delta 0.0179846750175	[-61.356 -51.465 -26.787	1.013	67.484	0.	0.	0.
0.]						

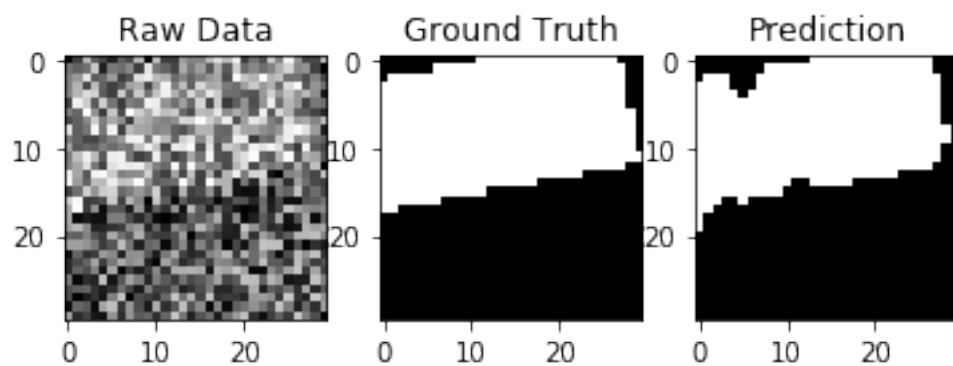
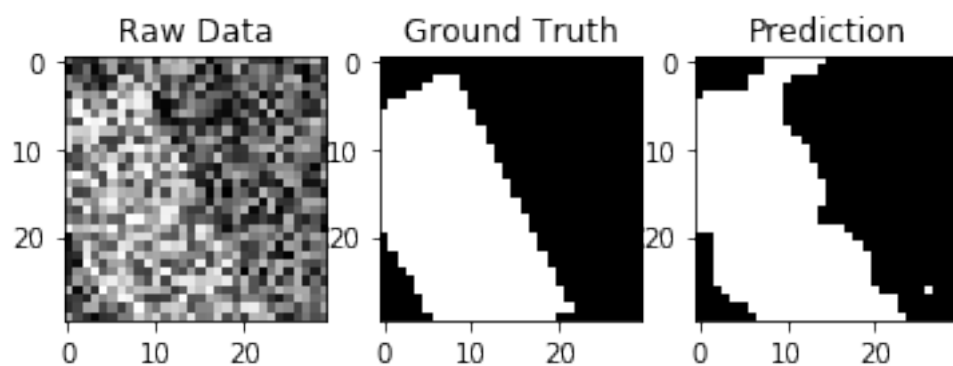
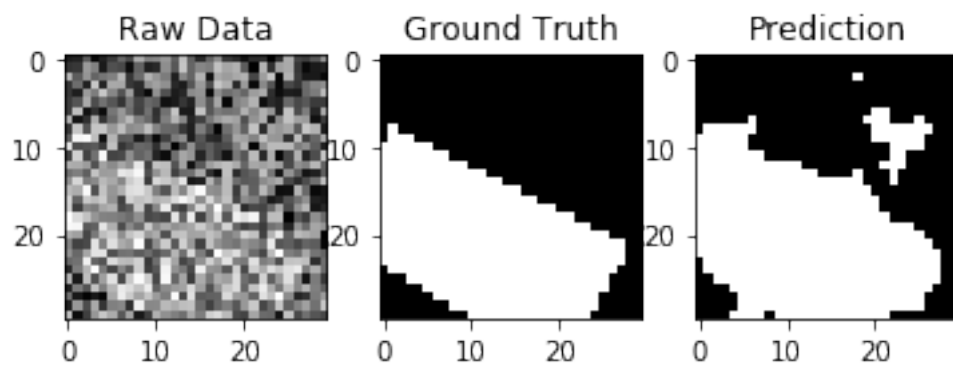
iter 31	delta 0.0230863110084	[-61.351 -51.459 -26.783	1.016	67.481	0.	0.	0.
0.]						
iter 32	delta 0.0168946947135	[-61.348 -51.456 -26.781	1.017	67.471	0.	0.	0.
0.]						
iter 33	delta 0.0217282927138	[-61.343 -51.451 -26.777	1.02	67.468	0.	0.	0.
0.]						
iter 34	delta 0.015929283587	[-61.34 -51.448 -26.775	1.021	67.46	0.	0.	0.
0.]						
iter 35	delta 0.0179949061733	[-61.337 -51.444 -26.772	1.023	67.454	0.	0.	0.
0.]						
iter 36	delta 0.0175085573578	[-61.333 -51.44 -26.77	1.025	67.449	0.	0.	0.
0.]						
iter 37	delta 0.0170478058484	[-61.33 -51.437 -26.767	1.027	67.444	0.	0.	0.
0.]						
iter 38	delta 0.0166106826215	[-61.326 -51.433 -26.764	1.029	67.439	0.	0.	0.
0.]						
iter 39	delta 0.016195415556	[-61.323 -51.429 -26.762	1.03	67.434	0.	0.	0.
0.]						
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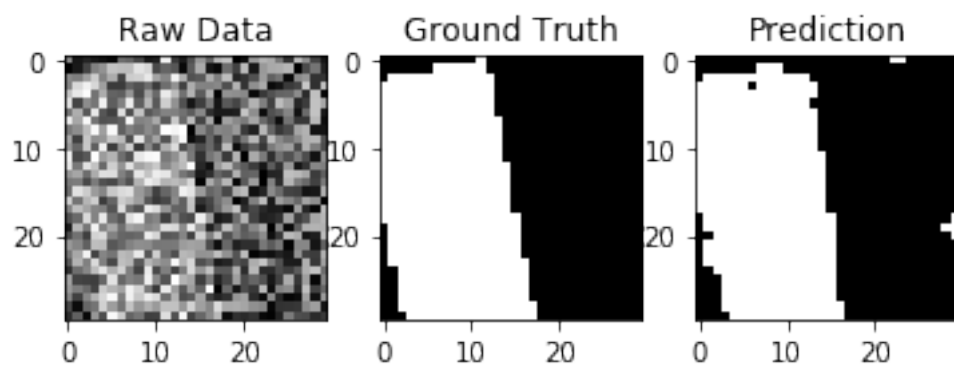
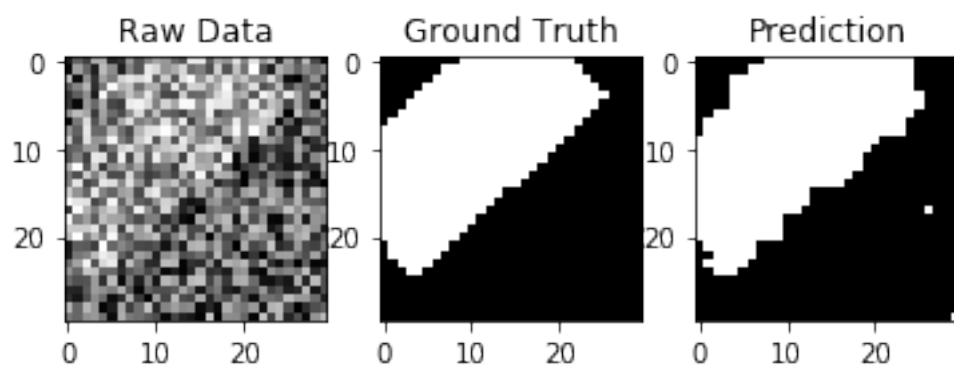
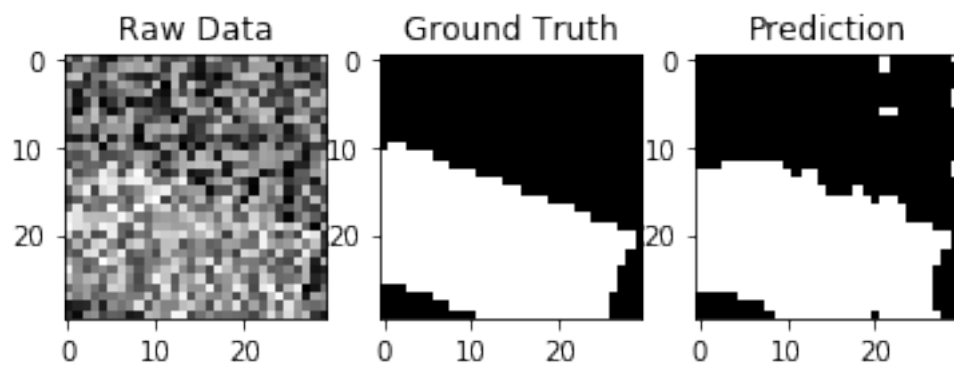
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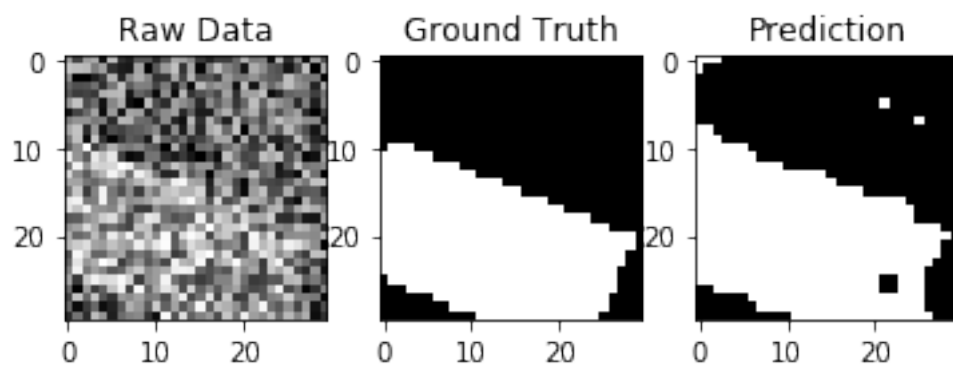
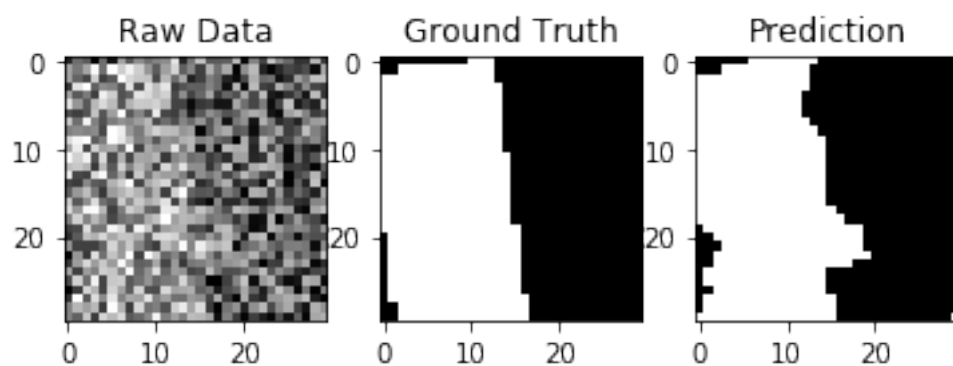
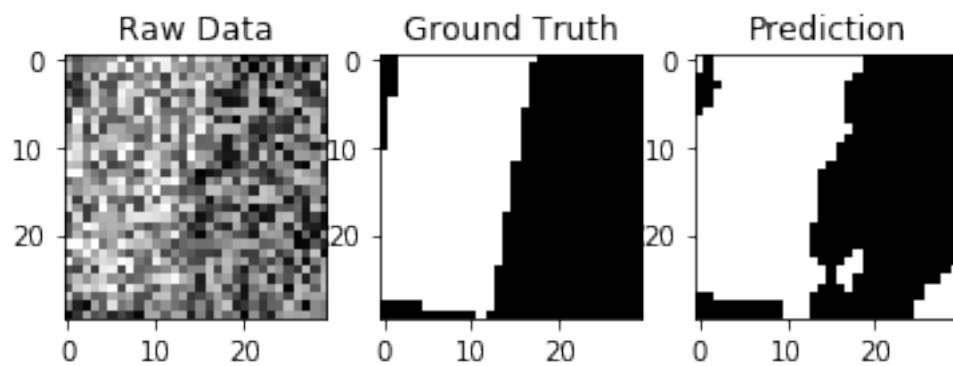
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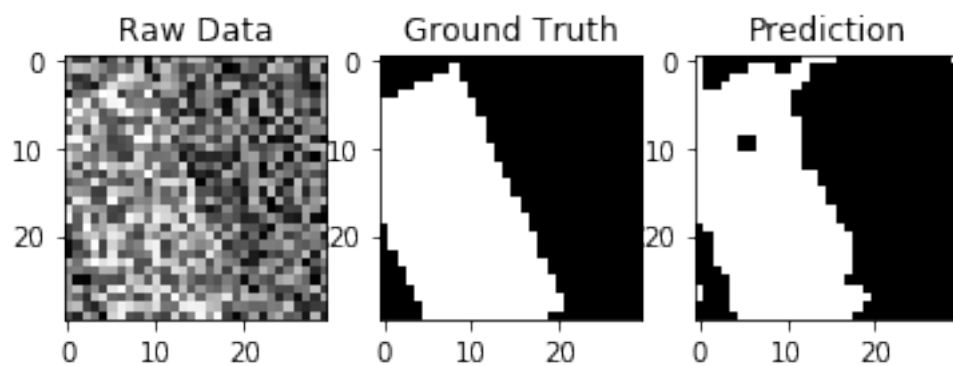
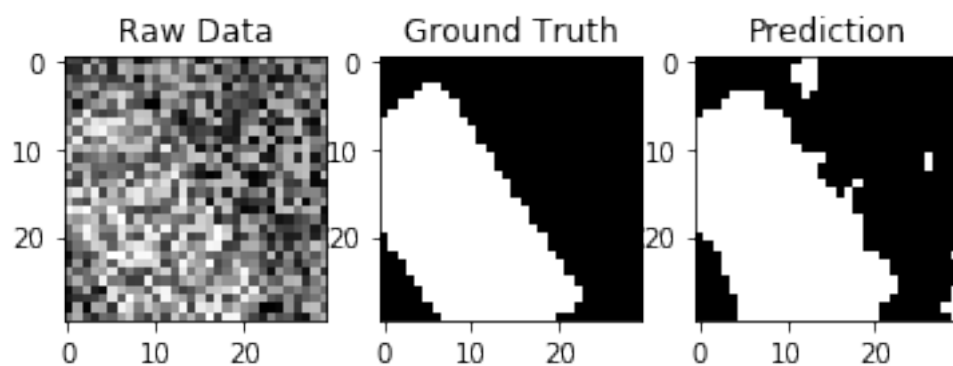
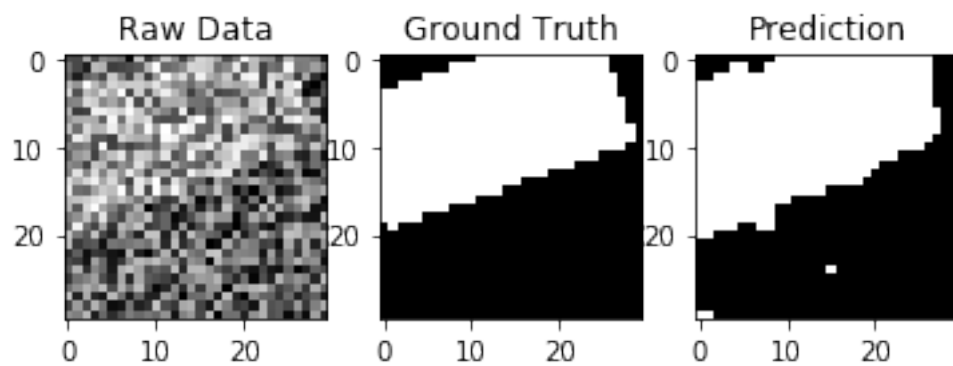


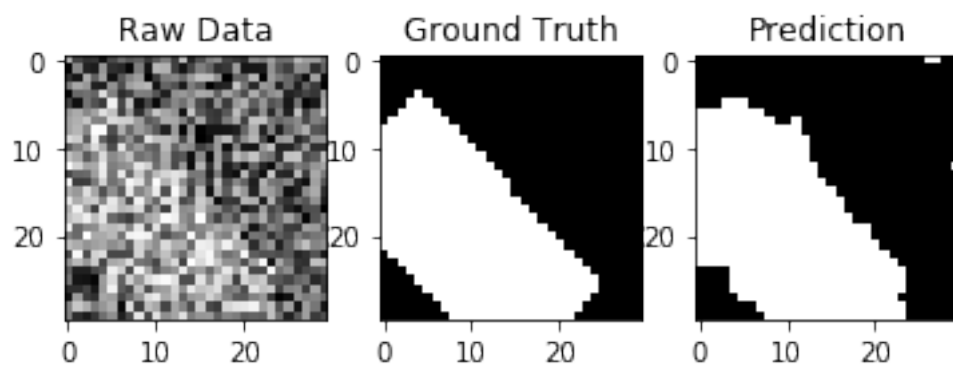
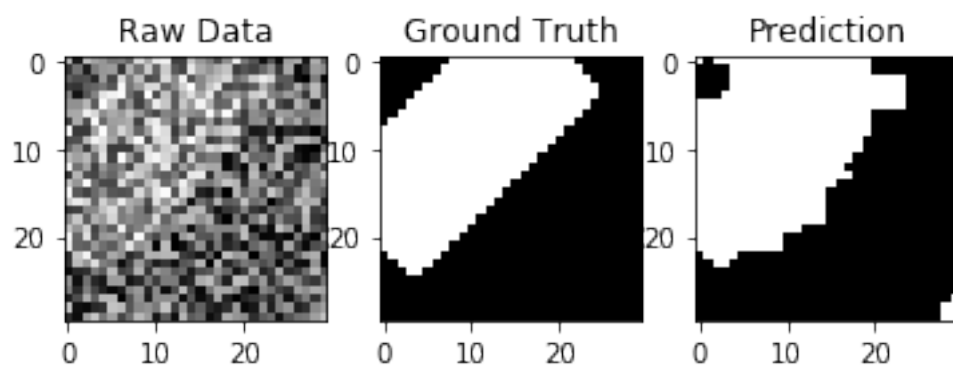
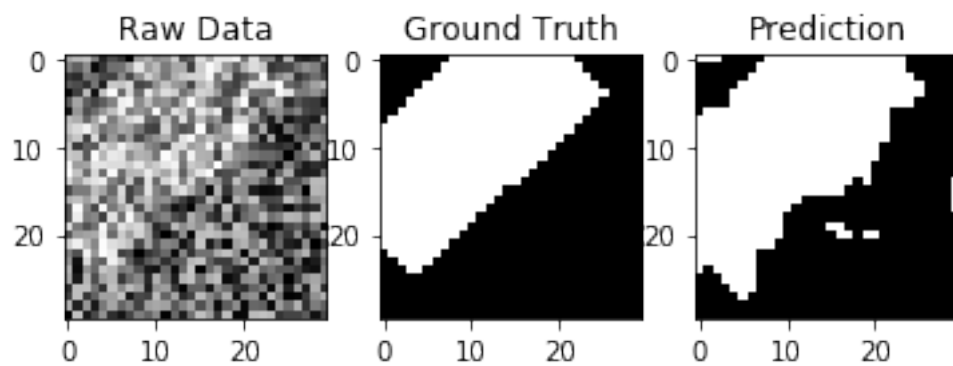


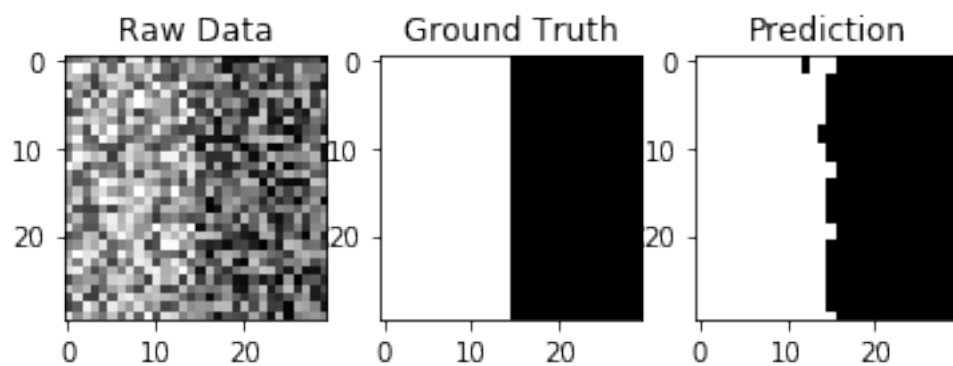
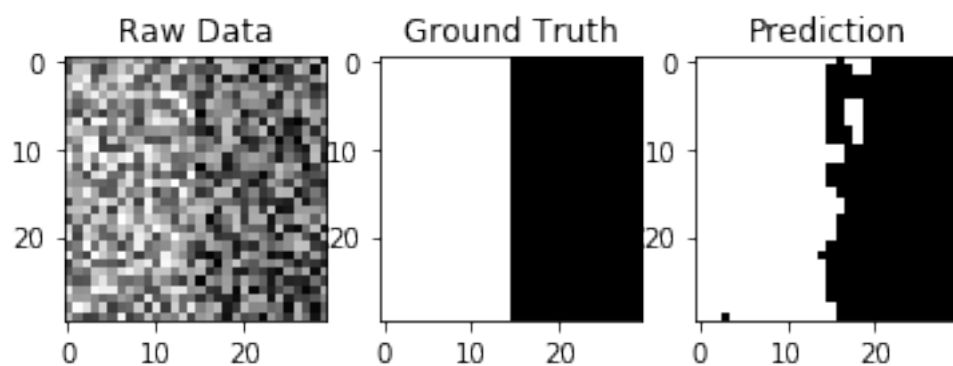
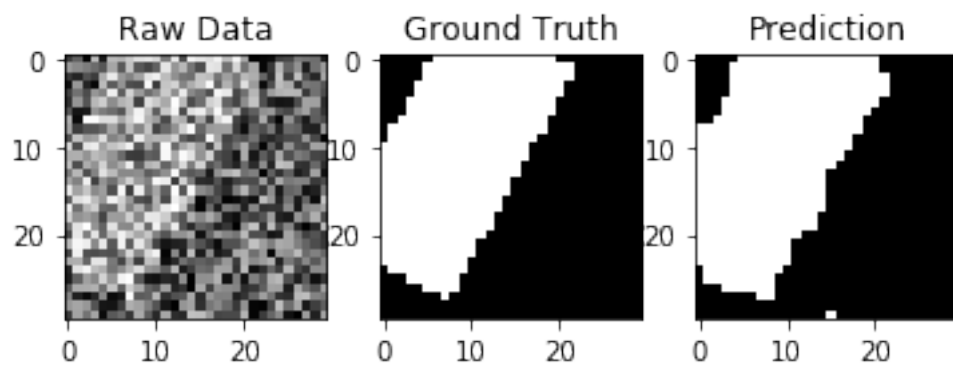


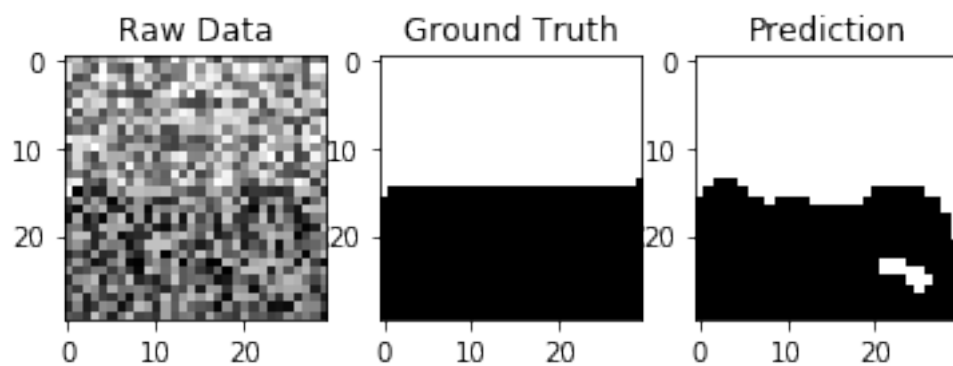
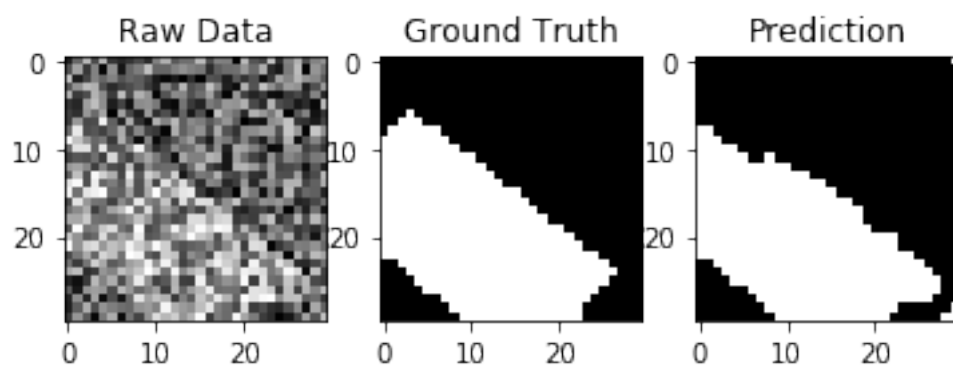
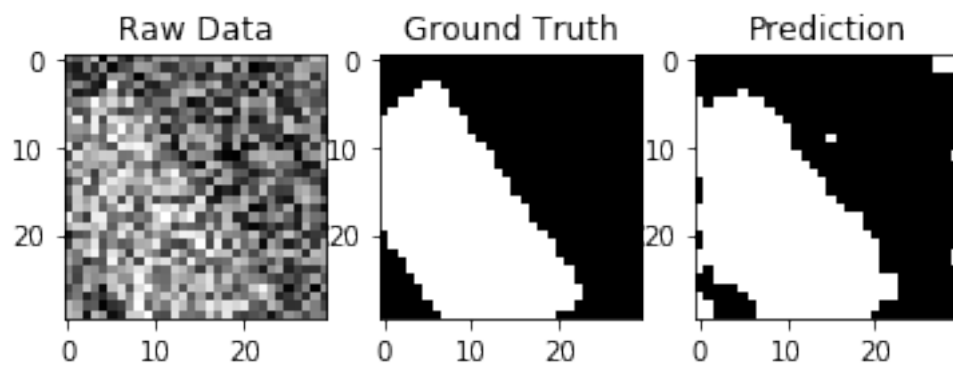


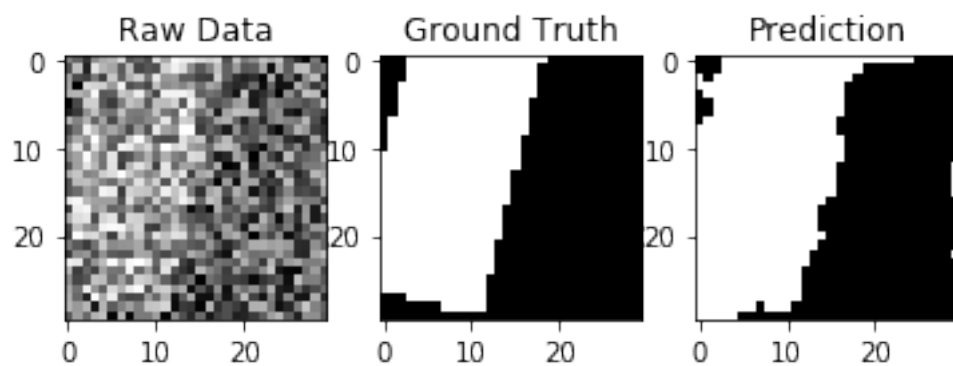
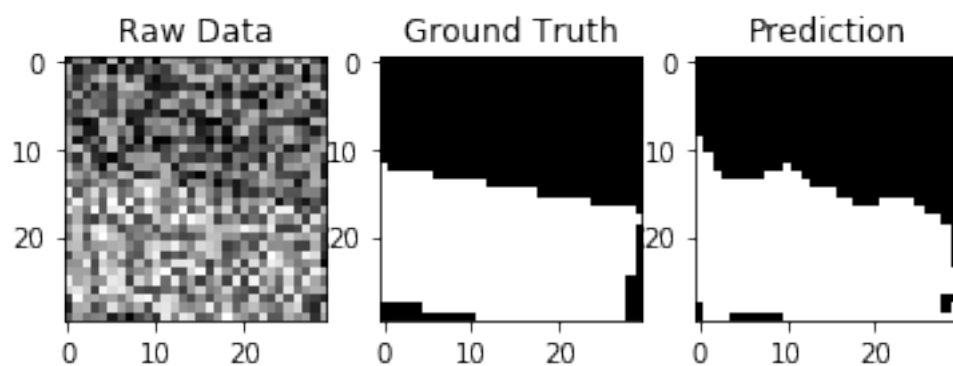
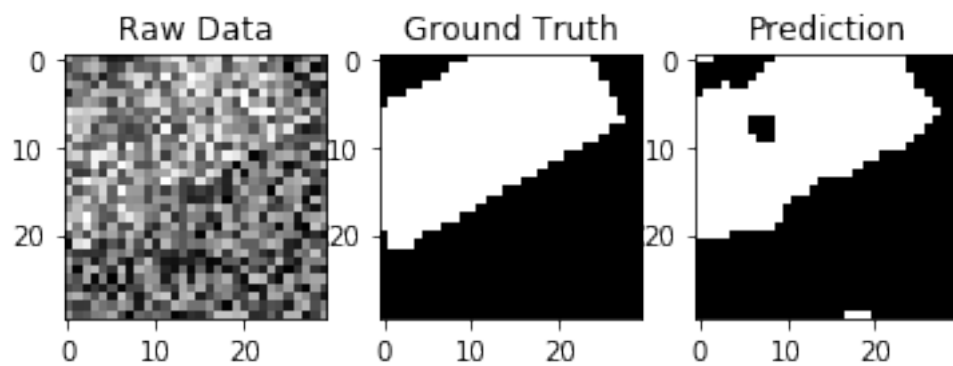


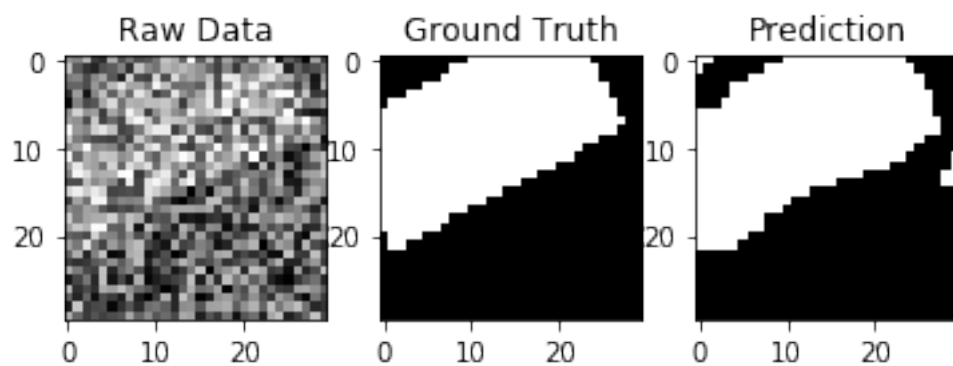
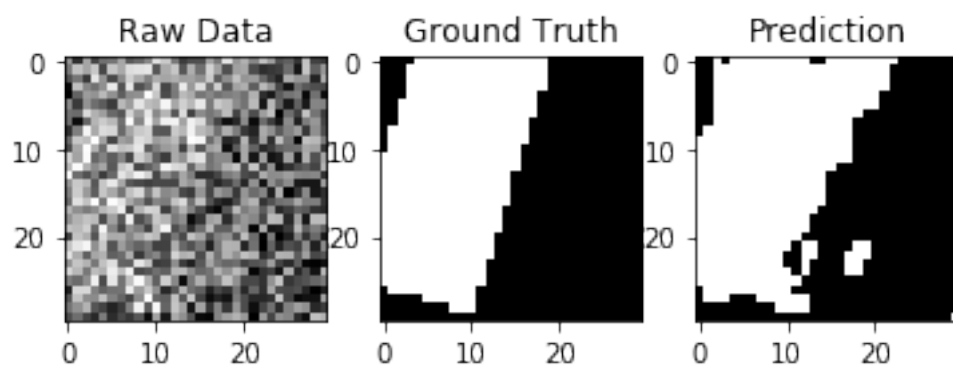
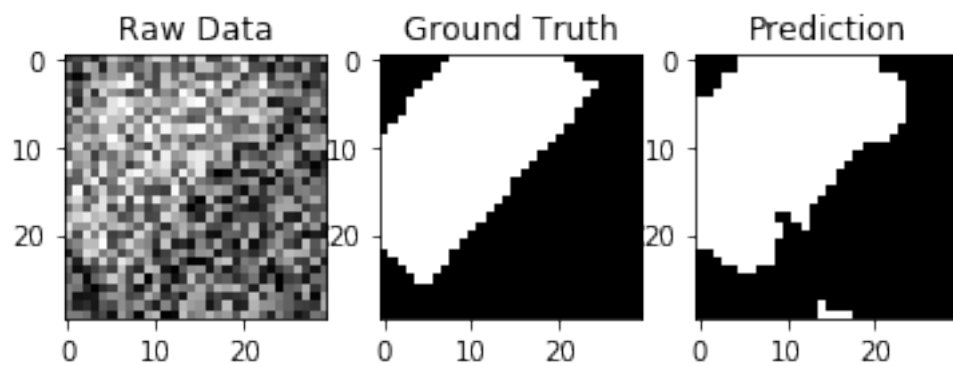


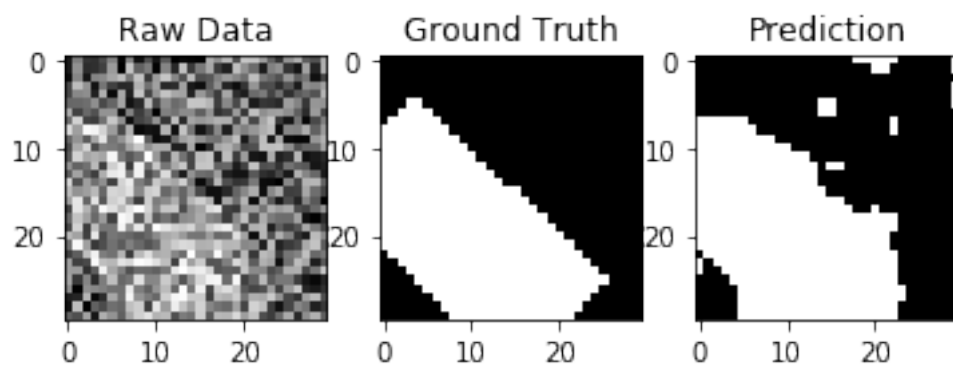
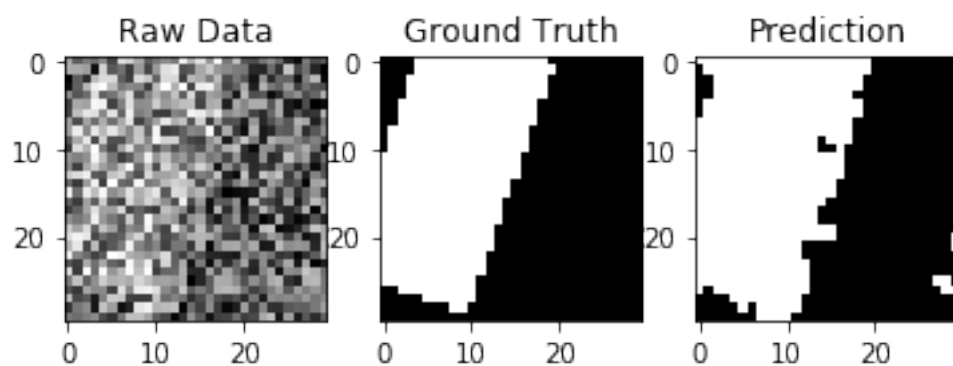
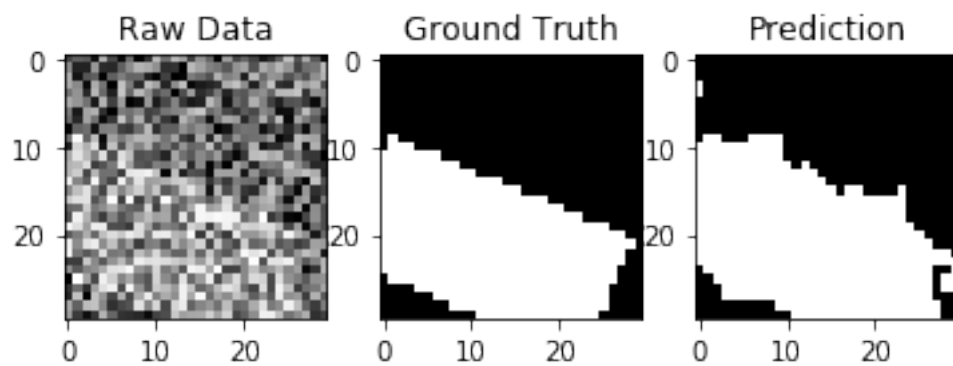


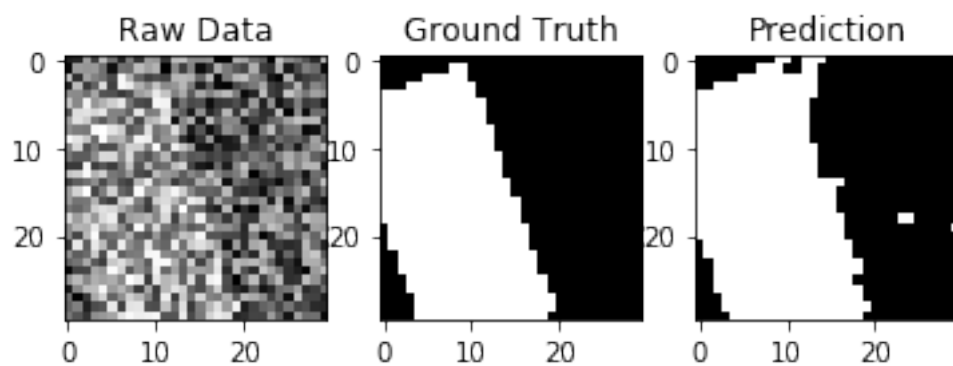
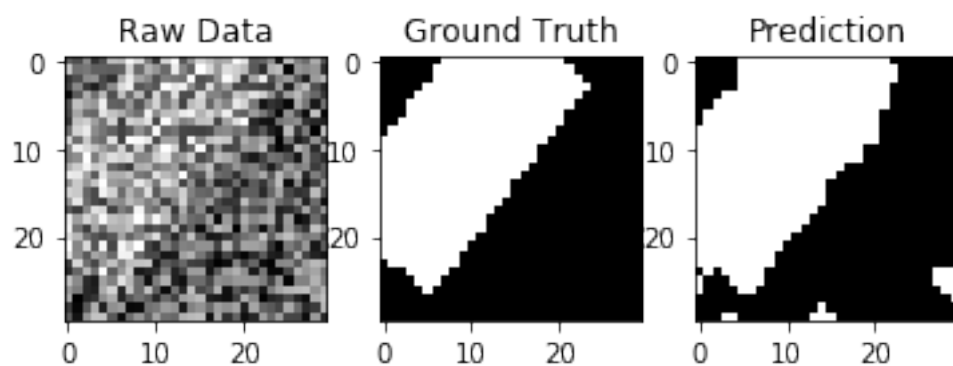
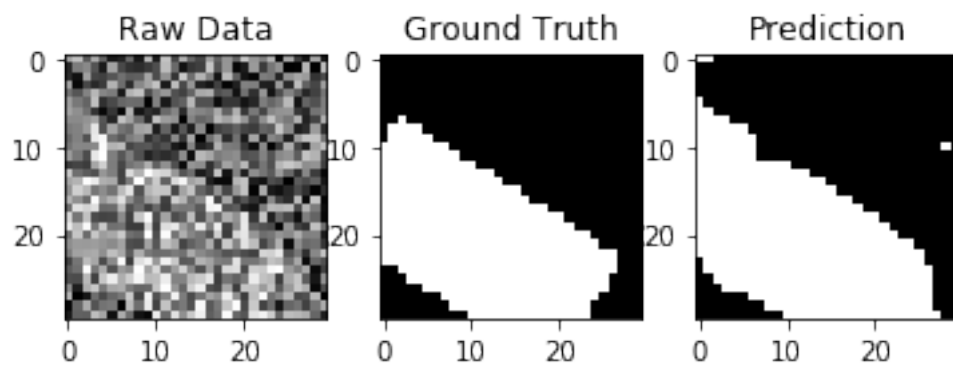


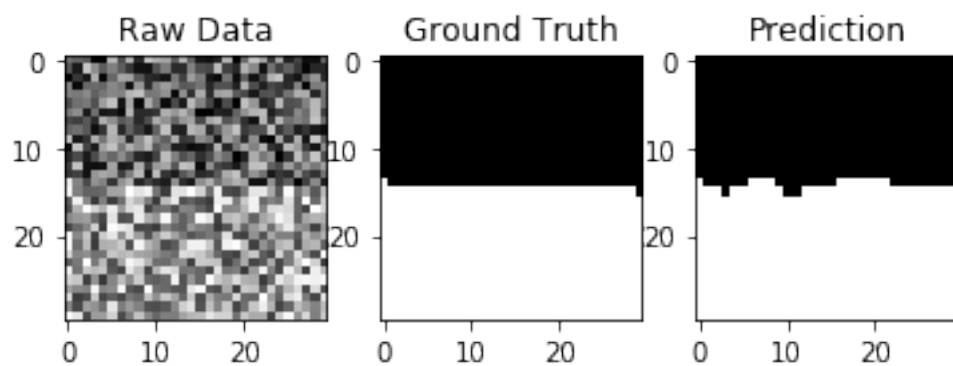
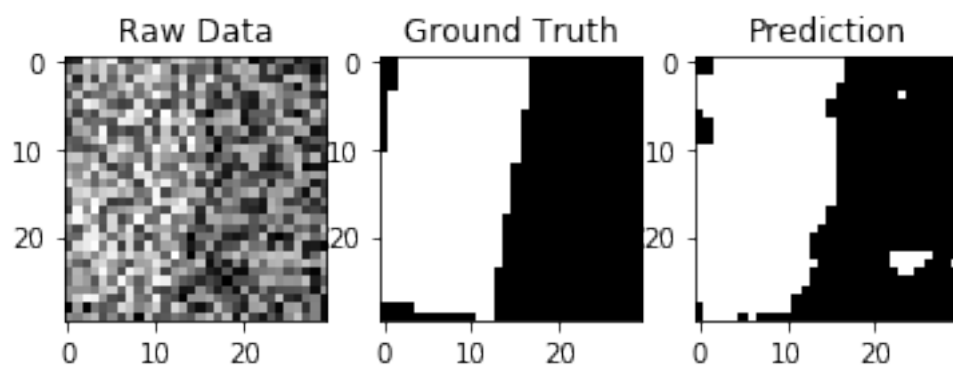
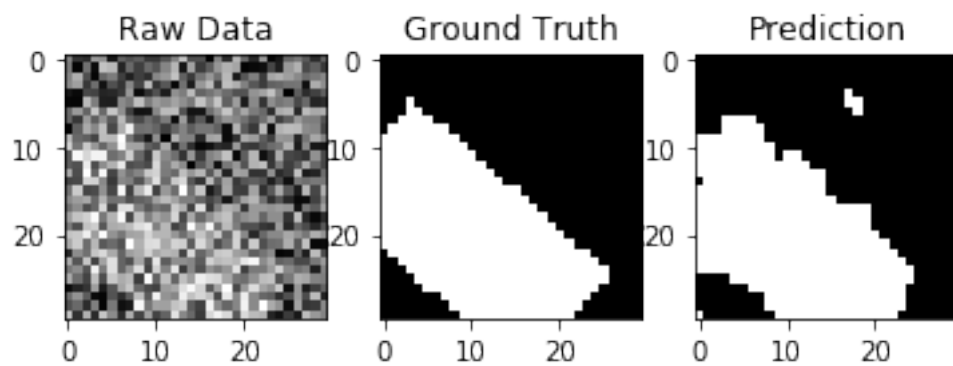


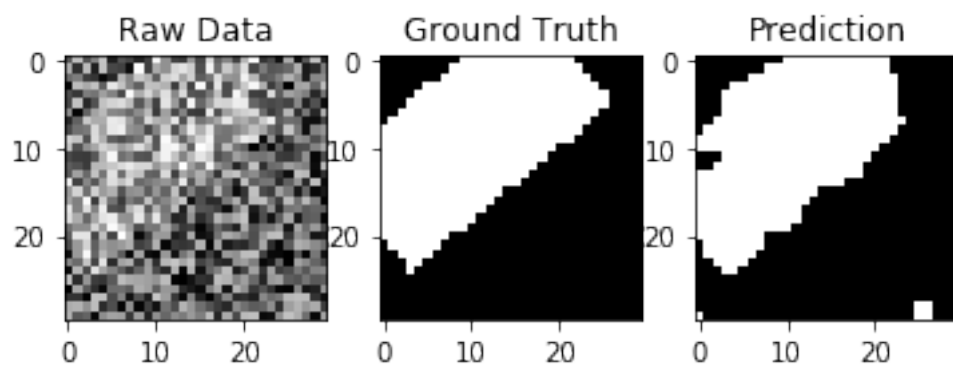
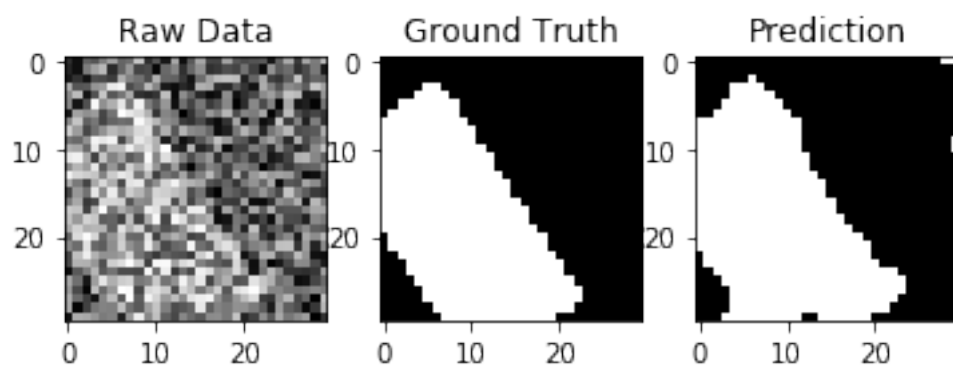
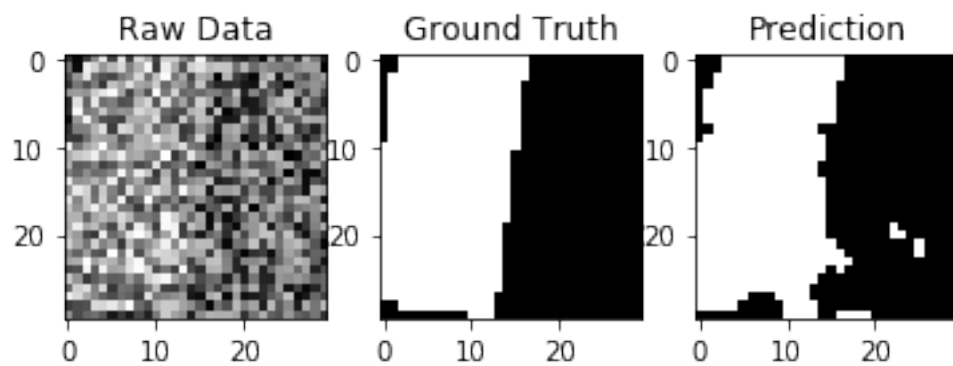


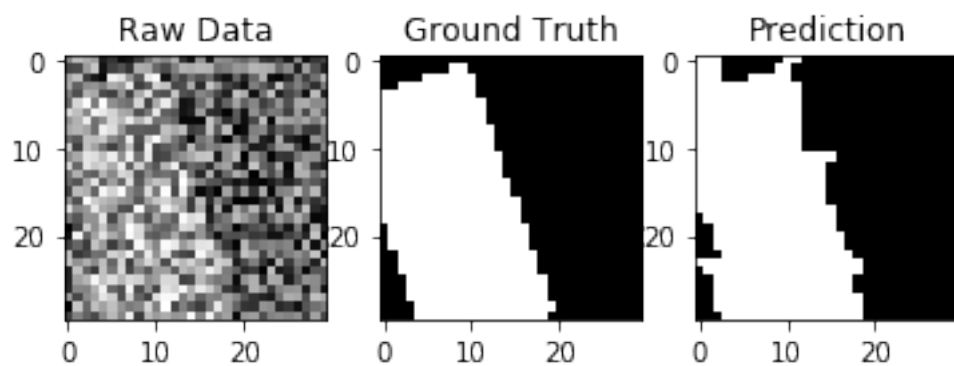
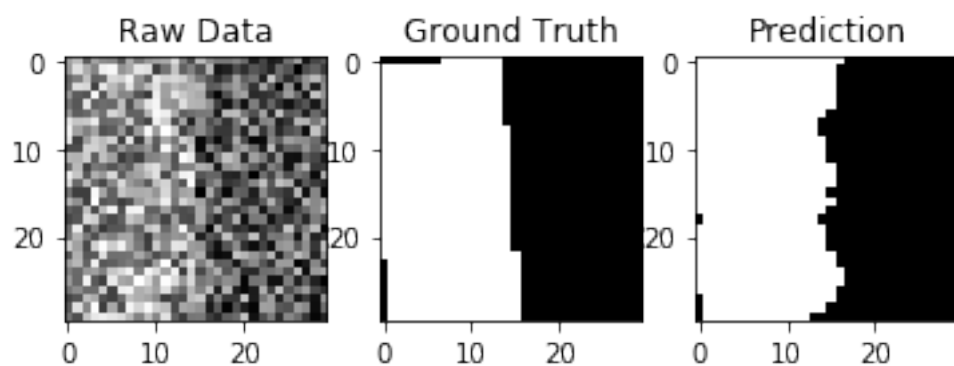
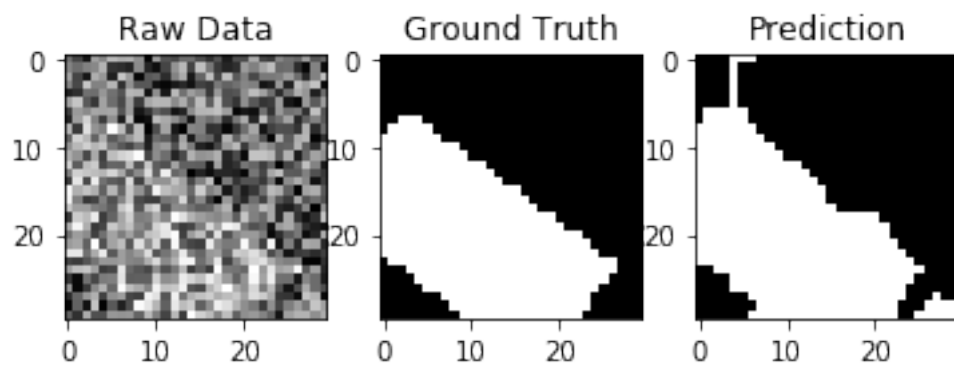


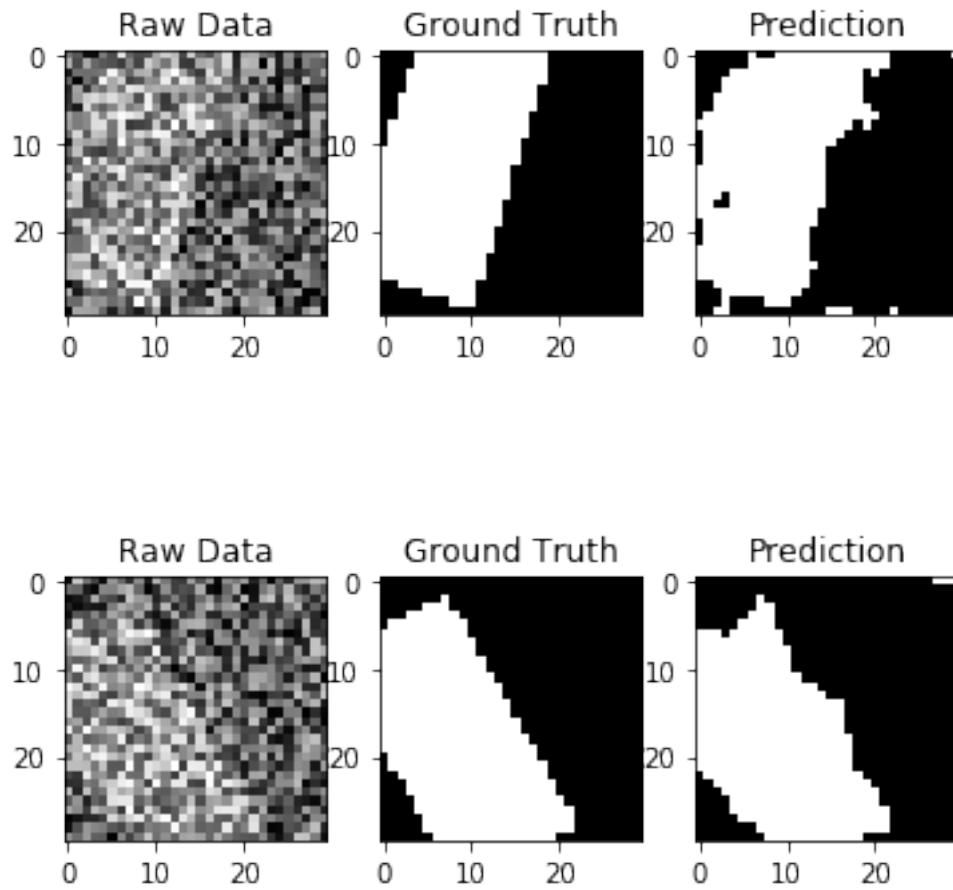












Yep, Graph cut works way better than ICM.

In []: